Mode of data elicitation, acquisition and response to surveys: a systematic review

K Hood, M Robling, D Ingledew, D Gillespie, G Greene, R Ivins, I Russell, A Sayers, C Shaw and J Williams

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Mode of data elicitation, acquisition and response to surveys: a systematic review

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The research reported in this issue of the journal was commissioned by the National Coordinating Centre for Research Methodology (NCCRM), and was formally transferred to the HTA programme in April 2007 under the newly established NIHR Methodology Panel. The HTA programme project number is 06/91/07. The contractual start date was in June 2005. The draft report began editorial review in September 2010 and was accepted for publication in March 2011. The commissioning brief was devised by the NCCRM who specified the research question and study design. The authors have been wholly responsible for all data collection, analysis and interpretation, and for writing up their work. The HTA editors and publisher have tried to ensure the accuracy of the authors’ report and would like to thank the referees for their constructive comments on the draft document. However, they do not accept liability for damages or losses arising from material published in this report.

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Abstract

Mode of data elicitation, acquisition and response to surveys: a systematic review

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Background: Many studies in health sciences research rely on collecting participant-reported outcomes and attention is increasingly being paid to the mode of data collection. Consideration needs to be given to the validity of response via different modes and the impact that choice of mode might have on study conclusions.

Objectives: (1) To provide an overview of the theoretical models of survey response and how they relate to health research; (2) to review all studies comparing two modes of administration for subjective outcomes and assess the impact of mode of administration on response quality; (3) to explore the impact of findings for key identified health-related measures; and (4) to inform the analysis of multimode studies.

Data sources: A broad range of databases (for example EMBASE, PsychINFO, MEDLINE, EconLit, SPORTDiscus, etc.) were chosen to allow as comprehensive a selection as possible, and they were searched up until the end of 2004.

Review methods: The abstracts were reviewed against inclusion/exclusion criteria. Full papers were retrieved for all selected abstracts and then screened again using more detailed inclusion criteria related to the measures used. Papers that were still included were reviewed in full and detailed data extracted. At each stage, abstracts or papers were reviewed by a single reviewer.

Results: The search strategy identified 39,253 unique references, of which 2156 were considered as full papers, with 381 finally included in the review. Two features of mode were clearly associated with bias in response; however, none of the features of mode was associated with changes in precision. How the measure was administered, by an interviewer or by the person themselves, was highly significantly associated with bias ($p < 0.001$). A difference in sensory stimuli was also significant ($p = 0.03$). When both of these were present the average overall bias was < 1 point on a percentage scale. In terms of mediating factors, there was some suggestion that there was an interaction between both telephone and computer for data collection and date of publication, supporting the theory that differences disappear as new technologies become commonplace. Single-item measures were also related to greater degrees of bias than multi-item scales ($p = 0.01$). Individual analysis of the Short Form questionnaire-36 items and Minnesota Multiphasic Personality Inventory (MMPI) showed a varied pattern across the different subscales, with conflicting results between the two types of study. None of the MMPI measures used to
detect deviant responding showed a relationship with the mode features tested. The limits of agreement analysis showed how variable measures were between modes at an individual rather than a group mean level.

**Limitations:** The search strategy covered the period up to 2004, so any new and emerging technologies were not included. Not all potential mode features were tested and there was limited information on potential mediating factors.

**Conclusions:** Researchers need to be aware of the different mode features that could have an impact on their results when selecting a mode of data collection for subjective outcomes. Further mode comparison studies, which manipulate mode features and directly assess impact over time, would be beneficial.

**Funding:** The National Institute for Health Research Health Technology Assessment programme.
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Glossary

**Acquiescence**  A response bias whereby respondents simply agree with an attitudinal statement regardless of content.

**Optimising**  The process of carefully and comprehensively proceeding through all cognitive steps required when answering a survey question.

**Satisficing**  A strategy of providing a satisfactory response to a survey question without the respondent expending the intended cognitive effort. This may be due to incomplete or biased or absent retrieval and/or integration of information when responding.
## List of abbreviations

<table>
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<th>Abbreviation</th>
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<tr>
<td>ACASI</td>
<td>audio computer-assisted self-interview</td>
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<td>AUC</td>
<td>area under the curve</td>
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<tr>
<td>CAPI</td>
<td>computer-assisted personal interview</td>
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<tr>
<td>CASI</td>
<td>computer-assisted self-administered interview</td>
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<tr>
<td>CAT</td>
<td>computerised adaptive testing</td>
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<tr>
<td>CATI</td>
<td>computer-assisted telephone interview</td>
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<tr>
<td>CI</td>
<td>confidence interval</td>
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<tr>
<td>ES</td>
<td>effect size</td>
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<tr>
<td>HRQoL</td>
<td>health-related quality of life</td>
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<tr>
<td>ICC</td>
<td>intracluster correlation coefficient</td>
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<tr>
<td>IRT</td>
<td>item response theory</td>
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<tr>
<td>IVR</td>
<td>interactive voice response</td>
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<tr>
<td>MeSH</td>
<td>medical subject headings</td>
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<tr>
<td>MMPI</td>
<td>Minnesota Multiphasic Personality Inventory</td>
</tr>
<tr>
<td>PDA</td>
<td>personal digital assistant (handheld computer)</td>
</tr>
<tr>
<td>PRISMA</td>
<td>Preferred Reporting Items for Systematic Reviews and Meta-Analyses</td>
</tr>
<tr>
<td>PROM</td>
<td>patient-reported outcome measure</td>
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<tr>
<td>QALY</td>
<td>quality-adjusted life-year</td>
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<tr>
<td>QoL</td>
<td>quality of life</td>
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<tr>
<td>RCT</td>
<td>randomised controlled trial</td>
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<tr>
<td>ROC</td>
<td>receiver operating characteristic</td>
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<tr>
<td>SAQ</td>
<td>self-administered questionnaire</td>
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<tr>
<td>SD</td>
<td>standard deviation</td>
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<tr>
<td>SF-36</td>
<td>Short Form questionnaire-36 items</td>
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All abbreviations that have been used in this report are listed here unless the abbreviation is well known (e.g. NHS), or it has been used only once, or it is a non-standard abbreviation used only in figures/tables/appendices, in which case the abbreviation is defined in the figure legend or in the notes at the end of the table.
Executive summary

Background

Many studies in health sciences research rely on collecting participant-reported outcomes. Although some of these are participant reports of factual information, such as adherence to drug regimes, that could be objectively validated, there is an increasing recognition of the importance of subjective measures such as attitude to, and perceptions of, health and services provision. Alongside the exponential increase in health-related literature devoted to participant-reported outcomes, attention is being paid to the method or mode of data collection. Much of this has been driven by the rapid development of new technologies, which can lead to increased ease, speed and efficiency of data capture alongside an increasing drive for maximising response rates. Survey methodologies (e.g. in the business, marketing, social and political sciences) have a literature base of their own, covering theory to practice, much of which has been only slowly recognised in the health arena. Few health-related outcome development papers indicate a theoretical approach to eliciting survey response and the focus in choosing a mode for a study is often based predominantly on improving response rates and minimising cost. The impact on the validity of response is not generally a consideration. In addition to this, in order to gain as complete a data set as possible, many studies are using multiple modes either to enhance participants’ choice (e.g. opting for web- or paper-based surveys) or to improve follow-up rates (e.g. non-responders getting telephone data collection). Although for practical reasons these choices are entirely justifiable, consideration needs to be given to the validity of response via different modes and the impact that the choice of mode or modes might have on the conclusions from a study.

Objectives

- To provide an overview of the theoretical models of survey response and how they relate to health research.
- To review all studies comparing two modes of administration for subjective outcomes and assess the impact of mode of administration on response quality.
- To explore the impact of findings for key identified health-related measures.
- To create an accessible resource for health science researchers, which will advise on the impact of the selection of different modes of data collection on response.
- To inform the analysis of multimode studies.

Methods

In order to inform the systematic review of mode comparison studies, a review of the theoretical models and how they relate to the health domain was undertaken. This clarified the need to focus on features of mode rather than crude modes per se in order to understand the way in which responses to subjective outcomes could be affected. From this, a theoretical model based on Tourangeau was proposed with four main features: administration (interviewer or self), use of the telephone, use of the computer and sensory stimuli (audio, visual or both). Additional features were proposed that may belong in a model of response as well as potential mediating factors, such as cognitive challenge of questions. This approach was used to define the data extraction and coding classifications for studies.
Owing to the large body of literature relating to survey methodology which is published outside the health research arena, all studies that incorporate a mode comparison were included, regardless of setting. This led to a broad search strategy covering a wide range of disciplines. In order to target methodological studies, some innovations in search strategy that separate out the process from traditional reviews of the effectiveness of interventions were undertaken.

**Identifying the literature**

For a study to be included in the review it needed to:

1. provide evidence of a comparison between two modes of data collection of either the same question or the same set of questions referring to the same theoretical construct
2. compare a construct that is subjective and cannot be externally validated
3. explicitly reference a comparison in the analysis
4. collect quantitative data, i.e. use structured questions and answers.

Studies were excluded from the review if they involved:

1. a comparison between a quantitative measure and one or more qualitative data collection methods/analyses (e.g. unstructured interviews, focus groups)
2. a comparator derived from routine clinical records – unless explicit reference to specific self-reported construct is made within those records
3. a comparison between the response of two different judges, i.e. comparing a response from an individual to that made by someone other than the responder, for example a clinician providing a diagnosis.

A broad range of databases (for example EMBASE, PsychINFO, MEDLINE, EconLit, SPORTDiscus, etc.) were searched with no restrictions on start date or language. Searches were conducted up until the end of 2004. A matrix-based research strategy was developed and tested, searching for combinations of terms that would imply a mode comparison study.

**Review process**

The abstracts (and titles only for some foreign-language papers with no English abstract) were reviewed against the inclusion/exclusion criteria. Full papers were retrieved for all selected abstracts and then screened again using more detailed inclusion criteria related to the measures used. Papers that were still included were reviewed in full and detailed data extracted. At each stage, abstracts or papers were reviewed by a single reviewer after a period of training. Training for each stage included an assessment of reliability and sensitivity.

In order to assess the quality of the evidence contributing to this review, each paper was assessed for methodological quality. Assessing the quality of evidence becomes particularly challenging in reviews of studies having diverse methodologies. In this particular review, randomised controlled trials were not necessarily expected and so a more generic quality assessment tool was needed. A new tool was developed from two existing tools and tested.

**Evidence synthesis**

An overview of the studies identified is presented descriptively, highlighting the different mode features identified in the theory review. Those with appropriate data are subjected to quantitative methods of synthesis using exploratory metaregression to identify the association between mode features and differences in response. The primary analysis is based on three key summary statistics calculated for each comparison. These are the absolute difference between the means
(standardised) of the two modes, the ratio of the largest to the smallest variance of the two modes and the effect size (ES; absolute mean difference/standard deviation) between two modes.

Between- and within-subject studies were analysed together, controlling for the study design. Analysis was conducted at two levels to account for clustering of comparisons within a study. This allowed for study-level characteristics, measure characteristics and mode features to be considered in a single model. The modelling approach assessed the four main mode features from the theoretical review, then tested the addition of other candidate features and then assessed model fit including other possible moderators of effect and identified interaction.

The two most frequently occurring outcomes – the Short Form questionnaire-36 items (SF-36) and the Minnesota Multiphasic Personality Inventory (MMPI) – are analysed in more depth using Mantel–Haenszel for between-group studies and Bland and Altman limits of agreements for within-group studies.

**Results**

The search strategy identified 39,253 unique references, of which 2156 were considered as full papers. Of these, 597 progressed to data extraction, with 381 finally included in the review. The most common reason (44%) for exclusion once the full paper was considered was that there was no actual mode comparison in the study. The majority of included studies were from North America (62%), with only 10% being from the UK.

Study designs were relatively evenly divided into between- and within-person studies (52% and 47%, respectively), with only 39% using some form of randomisation (random allocation for between-person studies and random ordering for within-person studies). In terms of quality assessment, most studies described their hypotheses and study design well, and drew appropriate conclusions (89%, 83% and 81% – good, respectively), but the description of participants, group allocation, potential impact of timing of data collection and presenting of variances was less good (22%, 50%, 27% and 35% – poor, respectively).

The 381 studies provided descriptions on 1282 outcome measures, of which 57% were health related. The most frequently reported outcomes were the SF-36 (17 studies) and the MMPI (9 studies). Thirty per cent of studies considered only a single outcome in their mode comparison, but most considered more (ranging from 1 to 21 outcomes). These studies also described a number of mode comparisons, giving in total 1522 comparisons between modes on multiple outcomes for analysis. Of these, 977 reported enough data to be included in the analysis of absolute mean differences, 910 in the analysis of the ratio of variances and 912 in the analysis of the ES.

Two features of mode were clearly associated with bias in response; however, none of the features of mode was associated with changes in precision. How the measure was administered, by an interviewer or by the person themselves, was highly significantly associated with bias ($p < 0.001$). A difference in sensory stimuli was also significant ($p = 0.03$). When both of these were present the average overall bias was < 1 point on a percentage scale. In terms of mediating factors, there was some suggestion that there was an interaction between both telephone and computer for data collection and date of publication, supporting the theory that differences disappear as new technologies become commonplace. Single-item measures were also related to greater degrees of bias than multi-item scales ($p = 0.01$).
Individual analysis of the SF-36 and MMPI showed a varied pattern across the different subscales, with conflicting results between the two types of study. None of the MMPI measures used to detect deviant responding showed a relationship with the mode features tested. The limits of agreement analysis showed how variable measures were between modes at an individual rather than at a group mean level.

**Conclusions**

**Implications for researchers**

Researchers need to be aware of the different mode features that could have an impact on their results when selecting a mode of data collection for subjective outcomes. If researchers use a mixture of modes within their study (commonly a change in mode if there is poor or non-response), then consideration needs to be given to ameliorating potential biases consequent on this and controlling for them in analysis.

The potential does exist for there to be simple correction factors developed; however, these are likely to be measure specific. In analysis of current mixed-mode studies, researchers cannot just assume that results are comparable where a difference in administration or sensory stimuli exists and they need either to undertake sensitivity analyses or to formally control for mode in the analysis.

**Recommendations for future research (in priority order)**

There are already numerous studies considering a large number of outcome measures. However, these need to be reported in a standardised way to allow researchers to be able to make informed decisions about choice of mode with a particular outcome in a population. The development of reporting standards akin to PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses), STROBE (Strengthening the Reporting of Observational Studies in Epidemiology) or CONSORT (Consolidated Standards of Reporting Trials) for mode comparison studies is urgently needed and could build on the quality assessment tool developed here.

Further mode comparison studies are required, but these need to be experimentally designed to manipulate mode features and directly assess the impact. This is preferable to more studies comparing two modes at a relatively pragmatic level without consideration of those features. Studies need to give consideration to evaluation and direct testing of the impact of some of the mediators of mode effects, as the lack of data presented in papers in this review limited our ability to analyse this component.

Further primary studies need to be done to evaluate the impact of mode features over time. There was a suggestion across studies that this occurred for ‘new’ technologies for data collection (telephone and computer), but the ‘learning effect’ for any mode over time will be important to evaluate further in order to inform studies with long-term follow-up over multiple time points. The potential biasing impact of this ‘learning effect’ over time could be seen in single-mode studies as well as mixed-mode ones.

The focus of this review has been on measurement for research purposes and, therefore, has focused predominantly on the impact of mode features on estimated effects at a group level. However, the increasing use of subjective patient-reported outcomes in clinical practice means that considerable further work is required to consider measurement equivalence and reliability of assessment for individuals rather than groups.
Funding

Funding for this study was provided by the Health Technology Assessment programme of the National Institute for Health Research.
Chapter 1
Introduction

Many studies in health sciences research rely on collecting participant-reported outcomes of some form or another. Although some of these are participant reports of factual information, such as adherence to drug regimes, that could be objectively validated, there is increasing recognition of the importance of subjective measures, such as attitude to, and perceptions of, health and services provision. In addition to this, measures relating to health status which are not objectively measurable, such as quality of life (QoL), are becoming key secondary or even primary outcomes in many studies. This has led to a rapid growth in the development and validation of such measures. Few clinical trials, even with interventions pharmacological or surgical in nature, would be run today without measuring the patients’ QoL and assessing the acceptability of the intervention being trialled. The US Food and Drug Administration has recognised the importance of the inclusion of such measures as QoL for registration purposes and the National Institute for Health and Clinical Excellence incorporates quality-adjusted life-years (QALYs) as part of its decision-making process.

Alongside the exponential increase in health-related literature devoted to participant-reported outcomes (such as QoL), attention is being paid to the method or mode of data collection. Much of this has been driven by two main components: the rapid development of new technologies that can lead to increased ease, speed and efficiency of data capture, alongside an increasing drive for maximising response rates. This has led to a wide variety of options for mode of data collection being available to the health science researcher, with some studies adopting multiple approaches to follow up as many of the participants as possible. Although this approach may make sense pragmatically, it needs to be informed by an understanding of the participant’s ability to respond and statistical adjustment for biases introduced by multimode usage.

Theoretical approach

Survey methodologies (e.g. in the business, marketing, social and political sciences) have a literature base of their own covering theory to practice, much of which has been only slowly recognised in the health arena. Few health-related outcome development papers indicate a theoretical approach to eliciting survey response.

Although theoretical approaches are rarely considered, there has been a focus on maximising data capture by improving response rates. Reviews have been conducted which consider how features of the survey instrument (e.g. presentation, length, incentives) impact on response rates. There has also been an increase in ways in which such data are collected – the mode of data collection. With increasing levels of technology, a wider variety of modes are in use. The main focus in choosing a mode for a study appears to be based predominantly on improving response rates and minimising cost. The impact on the validity of response is not generally a consideration. In addition to this, in order to gain as complete a dataset as possible, many studies are using mixed modes either to enhance participants’ choice (e.g. opting for web- or paper-based surveys) or to improve follow-up rates (i.e. non-responders getting telephone data collection). Although for practical reasons these choices are entirely justifiable, consideration needs to be given to the validity of response via different modes and the impact that choice of mode or modes might have on the conclusions from a study.
Psychological theories of survey response will be considered in depth in Chapter 2. However, survey non-response and increasing concerns about maintaining adequate levels of response have led researchers to seek to categorise different forms of non-response. For example, Groves and Couper distinguish non-response due to non-contact, refusal to co-operate and inability to participate. The use of incentives to maintain response has, in turn, fostered theoretical development about how such inducements work, which, for example, have focused upon economic theories of incentives through to models describing a broader consideration of social exchange. Comprehensive theories of survey involvement have also been introduced and tested empirically.

More recently, a paradigm shift has been described within survey methodology from a statistical model focused upon the consequences of surveying error to social scientific models exploring the causes of error. Attempts to develop such theories of (1) survey error, (2) decisions to participate and (3) response construction have been brought under the general banner of the Cognitive Aspects of Survey Methodology (CASM) movement. Understanding and reducing measurement error, rather than sampling error, is at the forefront of this endeavour. An impetus for recent theoretical developments is very much provided by technological innovation and diversity, and a requirement to understand the relative impact of different data collection modes upon survey response.

Several information-processing models describing how respondents answer questions have been proposed, which share a common core of four basic stages: (1) comprehension of the question; (2) retrieval of information from autobiographical memory; (3) use of heuristic and decision processes to estimate an answer; and (4) response formulation. These models describe mostly sequential processing. A good example of a sequential information processing model is provided by Tourangeau et al. For each stage, there are associated processes identified, which a respondent may or may not use when answering a question. Each stage and each process may be a source of response error.

As indicated above, there has been a substantial expansion in the modes of data elicitation and collection available to survey researchers over the last 30 years. In 1996, Tourangeau and Smith identified six methods that may be used. A quick look at the literature since then will show that this expansion has continued with measures utilised that include personal digital assistants (PDAs) and websites. Subsequently, Tourangeau et al. delineated 13 different modes of survey data collection (including remote data collection methods such as telephone, mail, e-mail and the internet), which they considered differed in terms of five characteristics: (1) how respondents were contacted; (2) the presentational medium (e.g. paper or electronic); (3) method of administration (via interviewer or self-administered); (4) sensory input channel used; and (5) response mode.

Variations even within the same mode of data collection further complicate comparison. For example, Honaker describes computer-administered versions of the Minnesota Multiphasic Personality Inventory (MMPI), which differ in terms of type of computer being used, different computer–user interfaces with inconsistent item presentation and response formats. Therefore, different computerised versions of a test cannot be easily generalised to other versions. Other variables that could mediate the effect of different modes of data collection have also been considered, including the overall pace of the interview, the order of survey item processing and the role of different mental models used by respondents. Although the role of different mental models used by respondents, in particular, is rarely assessed, it has been considered a potentially significant mediator of response behaviour.
The challenge for health sciences research

As described above, the first characteristic underlying the different modes of data collection considered by Tourangeau et al. was method of contact. Work assessing the impact of an integrated process of respondent approach, consent and data collection has addressed bias due to selective non-ascertainment (i.e. the exclusion of particular subgroups). This may be clearly identifiable subgroups, in terms of people without telephones or computers (for telephone or internet approaches), or less clearly identifiable subgroups, i.e. those with lower levels of literacy or the elderly (for paper-based approaches). There is also considerable work on improving response rates and the biases induced by certain subgroups being less likely to consent to take part in a survey.

Furthermore, an important question in health services research is the use of data collection methods within prospective studies, where patients have already been recruited via another approach. This could be within a clinic or other health service setting rather than the survey instrument being the method of approach as well as data collection. Edwards et al. have recently updated a review of the literature (both health and non-health) to identify randomised trials of methods of improving response rates to postal questionnaires. Another review in health-related research has focused on the completeness of data collection and patterns of missing data, as well as response rates.

Guidance is needed not just about the ‘best’ method to use and most appropriate theoretical model of response, but also the consequence of combining data collected via different modes. For example, a common multimethod approach is when a second mode of data collection is used when the first has been unsuccessful (e.g. using telephone interview when there has been no response to a postal approach). Criteria for judging equivalence of the two approaches are therefore required. Honaker uses the concepts of psychometric equivalence and experiential equivalence. The former describes when the two forms produce results with equal mean scores, identical distribution and ranking of scores and agreement in how scores correlate with other variables. The latter deals with how two forms may differ in how they affect the psychometric and non-psychometric components of the response task.

In order to inform health services research, guidance is needed which quantifies the differences between modes of data collection and indicates which factors are associated with the magnitude of this difference. These could be contextual-based in terms of where the participant is when the information is completed (e.g. health setting, own home, work), content based in terms of questionnaire topic (e.g. attitudes to sexual behaviour) or population based (e.g. elderly). The factors identified by Tourangeau et al. also need to be tested across a wide range of modes and studies.

Aim

The aim of this project is to identify generalisable features affecting responses to the different modes of data collection relevant to health research from a systematic review of the literature.

Objectives

- To provide an overview of the theoretical models of survey response and how they relate to health research.
To review all studies comparing two modes of administration for subjective outcomes and assess the impact of mode of administration on response quality.

To explore the impact of findings for key identified health-related measures.

To create an accessible resource for health science researchers, which will advise on the impact of the selection of different modes of data collection on response.

To inform the analysis of multimode studies.
Chapter 2

Theoretical perspectives on data collection mode

Background

Understanding the unique experience of both users and providers of health services requires a broad range of suitably robust qualitative and quantitative methods. Both observational (e.g. epidemiological cohort) and interventional studies [e.g. randomised controlled trials (RCTs)] may collect data in a variety of ways, and often require self-report from study participants. Increasingly in clinical studies, clinical indicators and outcomes will form part of an assessment package in which patient lifestyle choices and behaviour, attitudes and satisfaction with health-care provision are a major focus. Health researchers need both to be reassured and to provide reassurance that the measurement tools available are fit for purpose across a wide range of contexts. This applies not only to the survey instrument itself, but also to the way it is delivered and responded to by the participant.

Options for collecting quantitative self-reported data have expanded substantially over the last 30 years, stimulated by technological advances in telephony and computing. The advent of remote data capture has led to the possibility of clinical trials being conducted over the internet. Concerns about survey non-response rates have also led researchers to innovate – resulting in greater diversity in data collection. Consequently, otherwise comparable studies may use different methods of data collection. Similarly, a single study using a sequential mixed-mode design may involve, for example, baseline data collection by self-completion questionnaire and follow-up by telephone interview. This has led to questions about the comparability of data collected by the different methods.

In this chapter we apply a conceptual framework to examine the differences generated by the use of different modes of data collection. Although there is considerable evidence about the effect of different data collection modes upon response rates, the chapter addresses the processes that may ultimately impact upon response quality. The framework draws upon an existing cognitive model of survey response by Tourangeau et al., which addresses how the impact of different data collection modes may be mediated by key variables. Furthermore, the chapter extends the focus of the model to highlight specific psychological response processes that may follow initial appraisal of survey stimulus. Although much of the empirical evidence for mode effects has been generated by research in other sectors, the relevance for health research will be explored. In doing so, other mediators of response will be highlighted.

It is important to clarify what lies outside the scope of the current review. Although mode of data collection can impact upon response rate as well as response content, that is not the focus of this report. Similarly, approaches that integrate modes of data collection within a study or synthesise data collected by varying modes across studies are addressed only in passing. Although these are important issues for health researchers, this review concentrates on how the mode of data collection affects the nature of the response provided by respondents, with a particular emphasis on research within the health sciences.
Variance attributable to measurement method rather than the intended construct being measured has been well recognised in the psychological literature and includes biases such as social desirability and acquiescence bias.20 This narrative review has been developed alongside the systematic literature review of mode effects in self-reported subjective outcomes presented in the subsequent chapters.21 The chapter highlights for researchers how different methods of collecting self-reported health data may introduce bias and how features of the context of data collection in a health setting such as patient role may modify such effects.

Modes and mode features

What are modes?

Early options for survey data collection were either face-to-face interview, mail or telephone. Evolution within each of these three modes led to developments such as computer-assisted personal interview (CAPI), web-delivered surveys and interactive voice response (IVR). Web-based and wireless technologies, such as mobile- and PDA-based telephony, have further stimulated the development of data collection methods and offer greater efficiency than traditional data collection methods, such as paper-based face to face interviews.22 Within and across each mode a range of options are now available and are likely to continue expanding.

A recent example of technologically enabled mode development is computerised adaptive testing (CAT). Approaches such as item response theory (IRT) modelling allow for survey respondents to receive differing sets of calibrated question items when measuring a common underlying construct [such as health-related quality of life (HRQoL)].23 Combined with technological advances, this allows for efficient individualised patient surveys through the use of computerised adaptive testing.24 In clinical populations, CAT may reduce response burden, increase sensitivity to clinically important changes and provide greater precision (reducing sample size requirements).23 Although IRT-driven CAT may be less beneficial where symptoms are being assessed by single survey items, more general computer-aided testing that mimics the normal clinical interview may be successfully used in combination with IRT-based CAT.25

What are the key features of different data collection modes?

The choice of mode has natural consequences for how questions are worded. Face-to-face interviews, for example, may use longer and more complex items, more adjectival scale descriptors and show cards.26 In contrast, telephone interviews are more likely to have shorter scales, use only end-point descriptors and are less able to use visual prompts, such as show cards. However, even when consistent question wording is maintained across modes there will still be variation in how the survey approach is appraised psychologically by respondents.

The inherent complexity of any one data collection approach (e.g. the individual characteristics of a single face-to-face interview paper-based survey) and increasing technological innovation means that trying to categorise all approaches as one or other mode may be too simplistic. Attention has therefore been focused upon survey design features that might influence response. Two recent models by Groves et al.18 and Tourangeau et al.8 illustrate this. Tourangeau identified five features: (1) how respondents were contacted (e.g. by post, in person); (2) the presentational medium (e.g. paper or electronic); (3) method of administration (interviewer- or self-administered); (4) sensory input channel (e.g. visual or aural); and (5) response mode (e.g. handwritten, keyboard, telephone).27 Groves et al.18 also distinguished five features: degree of interviewer involvement, level of interaction with respondent, degree of privacy, channels of communication (i.e. sensory modalities) and degree of technology.28 Although both models cover similar ground, Groves et al.18 place a greater emphasis upon the nature of the relationship between the respondent and the interviewer. Both models attempt to isolate the active ingredients
of survey mode. However, Groves et al. note that in practice differing combinations of features make generalisation difficult – reflected in their emphasis upon each individual feature being represented as a continuum. Although research on data collection methods has traditionally referred to as ‘mode’, given the complexity highlighted above, where appropriate we use the term ‘mode feature’ in this chapter.

**How mode features influence response quality**

Common to several information-processing models of how respondents answer survey questions there are four basic stages: (1) comprehension of the question; (2) retrieval of information from autobiographical memory; (3) use of heuristic and decision processes to estimate an answer; and (4) response formulation. At each stage, a respondent may use certain processes when answering a question, which may result in a response error.

Of the features that might vary across data collection method, Tourangeau et al. proposed four features that may be particularly influential in affecting response: (1) whether a survey schedule is self-administered or interviewer administered; (2) the use of a telephone; (3) computerisation; and (4) whether survey items are read by (or to) the respondent. Although this chapter focuses on differences between these broad mode features, there may still be considerable heterogeneity within each. For example, computerisation in the form of an individual web-delivered survey may apparently provide a standardised stimulus (i.e. overall package of features) to the respondent, but different hardware and software configurations for each user may violate this assumption.

Tourangeau et al. considered three variables to mediate the impact of mode feature: degree of impersonality, the sense of legitimacy engendered by the survey approach and the level of cognitive burden imposed. Both impersonality and legitimacy represent the respondent’s perceptions of the survey approach and instrument. Cognitive burden, impersonality and legitimacy are a function of the interaction between the data collection method and the individual respondent (and subject to individual variation). Nevertheless, the level of cognitive burden experienced by individuals is less dependent upon the respondent’s psychological appraisal of the survey task than perceptions of either impersonality or legitimacy.

The relationships among these mode features, mediating variables and three response quality indicators (rate of missing values, reliability and accuracy) are shown in Figure 1 and have been previously described by Tourangeau et al. In this chapter, we further distinguish between psychological appraisals and psychological responses. Psychological appraisals entail the initial processing of salient features by individual respondents and incorporate the mediators described by Tourangeau et al. Two additional appraisal processes are included (leverage-saliency and social exchange) and are described below. Initial appraisal then moves onto psychological response processes. In this amended model, these processes include optimising/satisficing, impression management and acquiescence. Each of these processes is described below and together they represent differing theoretical explanations for an individual’s response. The extent to which they are distinct or related processes is also examined.

Other features may also modify response and are added to the chapter framework. They include features of the ‘respondent’ (the information provider) and ‘construct’ (what is being measured). These features are not directly related to the method of data collection. Some of these features are implied by the mediators described by Tourangeau et al. (e.g. the sensitivity of the construct is implicit to the importance of ‘impersonality’). Nevertheless, we consider it important to separate out these features in this framework. Examples of both sets of features are provided, but are intended to be indicative rather than exhaustive listings. Finally, although there may be no
unique feature to distinguish between data collection in health and other research contexts, we have used, where we can, examples of particular relevance to health.

**How are data collection stimuli appraised by respondents?**

**Impersonality**

The need for approval may restrict disclosure of certain information. Static or dynamic cues (derived from an interviewer’s physical appearance or behaviour) provide a social context that may affect interaction. Self-administration provides privacy during data collection. Thus, Jones and Forrest found greater rates of reported abortion among women using self-administration methods than in personal interview. People may experience a greater degree of privacy when interacting with a computer and feel that computer-administered assessments are more anonymous.

The greater expected privacy for methods such as audio computer-assisted self-interview (ACASI) has been associated with increased reporting of sensitive and stigmatising behaviours. It is therefore possible that humanising a computerised data collection interface (e.g. the use of visual images of researchers within computerised forms) could increase misreporting. For example, Sproull et al. found higher social desirability scores among respondents to a human-like computer interface compared with a text-based interface. However, others have found little support for this effect in social surveys. Certain data collection methods may be introduced specifically to address privacy concerns – for example, IVR and telephone ACASI. However, there is also evidence that computers may reduce feelings of privacy. The need for privacy will vary with the sensitivity of the survey topic. Although Smith found the impact of the presence of others in response to the US General Social Survey to be mostly negligible, some significant effects were found. For example, respondents rated their health less positively when reporting in the presence of others than when lone respondents.

**Legitimacy**

Some methods restrict opportunities for establishing researcher credentials, for example when there is no interviewer physically present. A respondent’s perception of survey legitimacy could also be enhanced, albeit unintentionally, by the use of computers. Although this may be only a transient phenomenon, as computers become more familiar as data collection tools, other technological advances may produce similar effects (e.g. PDAs).
Cognitive burden
Burden may be influenced by self-administration, level of computerisation and the channel of presentation. Survey design that broadly accommodates the natural processes of responding to questions across these features is likely to be less prone to error.

Leverage–saliency theory
This general model of survey participation was proposed by Groves et al. and evaluates the balance of various attributes contributing to a decision to participate in a survey. Each attribute (e.g. a financial incentive) varies in importance (leverage) and momentary salience to an individual. Both leverage and salience may vary with the method of data collection and interact with other attributes of the survey (e.g. item sensitivity). Thus, face-to-face interviewers may be able to convey greater salience to responders through tailoring their initial encounter. This common thread of the presence of an interviewer may enhance the perceived importance of the survey to a respondent, which, first, may increase their likelihood of participating (response rate) and, second, enhance perceived legitimacy (response quality). The former effect – ‘participation decisions alone’ – is not examined further in this review. It is possible that the latter effect of mode feature on response quality may be particularly important in clinical studies if data are being collected by face-to-face interview with a research nurse, for example, rather than by a postal questionnaire.

Social exchange theory
This theory views the probability of an action being completed as dependent upon an individual's perception of the rewards gained and the costs incurred in complying, and his or her trust in the researcher. Dillman applied the theory to explaining response to survey requests – mostly in terms of response rate, rather than quality. However, he noted how switching between different modes within a single survey may allow greater opportunities for communicating greater rewards, lowering costs and increasing trust. This focus upon rewards may become increasingly important as response rates in general become more difficult to maintain. Furthermore, the use of a sequential mixed-mode design for non-respondent follow-up within a survey may enhance perceptions of the importance of the research itself by virtue of the researcher's continued effort.

Unlike the first three appraisal processes described above, both leverage–saliency and social exchange address broader participation decisions. Features of different data collection modes may affect such decision-making, for example through perceived legitimacy. Other features in the framework considered to modify response may also influence participation decisions according to these theories (e.g. the sensitivity of the construct being measured).

Explaining mode feature effects: psychological responses following appraisal
Initial appraisal of survey stimulus will result in a response process, which further mediates response quality. Several explanatory psychological theories have been proposed. We focus upon three general theories of response formulation (optimising/satisficing, social desirability and acquiescence).

‘Taking the easy way out’ – optimising and satisficing
Krosnick described ‘optimising’ and ‘satisficing’ as two ends of a continuum of thoroughness of the response process. Full engagement in survey response represents the ideal response strategy (optimising), in contrast to incomplete engagement (satisficing). The theory acknowledges the cognitive complexity of survey responding. A respondent may proceed through each cognitive step less diligently when providing a survey response or may omit information retrieval and judgement completely (examples of weak and strong satisficing, respectively). In either situation, respondents may use a variety of decision heuristics when responding. Three factors are considered to influence the likelihood of satisficing: respondent
ability, respondent motivation and task difficulty. Krosnick defines respondent ability (or cognitive sophistication) as the ability to retrieve information from memory and integrate it into verbally expressed judgements. Optimising occurs when respondents have sufficient cognitive sophistication to process the request, when they are sufficiently motivated and when the task requirements are minimal.

Mode feature effects may influence optimising through differences in non-verbal communication, interview pace (speed) and multitasking. First, the enthusiastic non-verbal behaviour of an interviewer may stimulate and maintain respondent motivation. Experienced interviewers react to non-verbal cues (e.g. expressions relating to lack of interest) and respond appropriately. Such advantages are lost in a telephone interview with interviewers relying on changes in verbal tones to judge respondent engagement. Although the role of an interviewer to enhance the legitimacy of the survey request was highlighted in Tourangeau et al.'s framework, this additional motivation and support function was not clarified. Second, interview pace may differ between telephone and face-to-face contact, in part because silent pauses are less comfortable on the telephone. A faster pace by the interviewer may increase the task difficulty (cognitive burden) and encourage respondents to take less effort when formulating their response. Pace can vary between self- and interviewer-administered methods. A postal questionnaire may be completed at respondents' own pace, allowing them greater understanding of survey questions compared with interviewer-driven methods. Tourangeau et al. omitted pace as a mediating variable from their model of mode effects because they considered that insufficient evidence had accrued to support its role. Interview pace has been suggested as an explanation for observed results, but the effects of pace have not necessarily been tested independently from other mode features (e.g. see Kelly et al.). Nevertheless, it is discussed here because of its hypothesised effect. Finally, distraction due to respondent multitasking may be more likely in telephone interviews than in face-to-face interviews (e.g. telephone respondents continuing to interact with family members or conduct household tasks while on the telephone). Such distraction increases task difficulty and thus may promote satisficing.

Optimising/satisficing has been used to explain a variety of survey phenomena, for example response order effects (where changes in response distributions result from changes in the presentational order of response options). Visual presentation of survey questions with categorical response options may allow greater time for processing initial options leading to primacy effects in those inclined to satisfice. Weak satisficing may also result from the termination of evaluative processing (of a list of response options) when a reasonable response option has been encountered. This may occur for response items with adjectival response scales and also for ranking tasks. In contrast, aural presentation of items may cause respondents to devote more effort to processing later response options (which remain in short-term memory after an interviewer pauses), leading to recency effects in satisficing respondents. Telephone interviews can increase satisficing (and social desirability response bias) compared with face-to-face interviews. An example of a theoretically driven experimental study that has applied this parsimonious model to studying mode feature effects is provided by Jäckle et al. In the setting of an interviewer-delivered social survey, they evaluated the impact of question stimulus (with or without show cards) and the physical presence or absence of interviewer (face to face or telephone). In this instance, detected mode feature effects were attributable not to satisficing, but to social desirability bias instead.

Social desirability
The tendency for individuals to present themselves in a socially desirable manner in the face of sensitive questions has long been inferred from discrepancies between behavioural self-report and documentary evidence. Response effects due to self-presentation are more likely when respondents' behaviour or attitudes differ from their perception of what is socially desirable.
This may result in over-reporting of some behaviours and under-reporting of others. Behavioural topics considered to induce over-reporting include being a good citizen and being well informed and cultured. Under-reporting may occur with certain illnesses (e.g. cancer and mental ill-health), illegal and non-normative behaviours and financial status. An important distinction has been made between intentional impression management (a conscious attempt to deceive) and unintentional self-deception (where the respondent is unaware of his or her behaviour). The former has been found to vary according to whether responses were public or anonymous, whereas the latter was invariant across conditions.

Most existing data syntheses of mode effects have related to social desirability bias (Table 1). Sudman and Bradburn indicated the importance of the method of administration upon socially desirable responding. They found a large difference between surveys either telephone- or self-administered compared with face-to-face interviews. Differences in social desirability between modes have been the subject of subsequent meta-analyses by de Leeuw, Richman et al., and Dwight and Feigelson. De Leeuw analysed 52 studies, conducted between 1947 and 1990, comparing telephone interviews, face-to-face interviews and postal questionnaires. There was no overall difference in socially desirable responding between face-to-face and telephone surveys among 14 comparisons. There was, however, more bias in telephone interviews in the nine studies published before 1980, but no difference in the later studies. There was less socially desirable responding in postal surveys than in both face-to-face surveys (13 comparisons, mean $r = 0.09$) and telephone surveys (five comparisons, mean $r = 0.06$). The presence of an interviewer (telephone or face to face), therefore, appears to determine socially desirable responding.

The review included both subjective and objective outcomes, and health issues were the most prominent topic covered.

The meta-analysis of Richman et al. compared computer-administered questionnaires, paper-and-pencil questionnaires and face-to-face interviews in 61 studies. Controlling for moderating factors, there was less social desirability bias in computer administration than in paper-and-pencil administration [effect size (ES) for difference of 0.39]. This advantage over paper-and-pencil methods was greater in studies conducted before 1975 (ES = 0.74), when responses were provided when alone (ES = 0.82) and when backtracking (i.e. ability to move back to earlier section of questionnaire) was available (ES = 0.65). However, when social desirability was inferred from other measures (rather than measured directly) there was more bias using computer administration controlling for moderators (ES = 0.46). Compared with face-to-face interviews, computer administration was associated with less bias overall (ES = 0.19). However, the opposite was true when the construct assessed was personality (ES = 0.73) and in more recently published studies (ES = 0.79).

Dwight and Feigelson compared impression management/self-deceptive enhancement in computer-administered measures and either paper-and-pencil or face-to-face measures. Less impression management bias was found for computer administration than for non-computer formats, but the difference was small (ES = −0.08). Individual study ESs reduced significantly over time, indicating a diminishing impact of computerisation. Dwight and Feigelson pointed to the recent positive ESs, which they felt was consistent with a ‘Big Brother syndrome’ – respondents fear monitoring and controlling by computers. There was no observed difference between data collection method on scores of self-deceptive enhancement.

It is worth commenting upon the methodological quality of these reviews. None provided an explicit search strategy, although all, apart from Sudman and Bradburn, described keywords. Dwight and Feigelson’s search was based upon an initial citation search, whereas only Richman et al.’s review provided explicit eligibility criteria for included studies. Sudman and Bradburn developed a comprehensive coding scheme that was later extended in de Leeuw’s review.
### TABLE 1 Reviews of mode effects in socially desirable responding

<table>
<thead>
<tr>
<th>Review details</th>
<th>Modes compared</th>
<th>No. of comparisons</th>
<th>Primary result</th>
<th>Evidence of effect moderators</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sudman and Bradburn</td>
<td>Face to face, self-administration</td>
<td></td>
<td>RE: face to face = 0.19, self-administration = 0.32</td>
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<tr>
<td>Years: not reported^a</td>
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<tr>
<td>Effect estimate:</td>
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<tr>
<td>relative RE</td>
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<tr>
<td>Studies: n = 305^a</td>
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<tr>
<td>1. Strong possibility of SD answer</td>
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<tr>
<td>2. Some possibility of SD answer</td>
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<tr>
<td>3. Little/no possibility of SD answer</td>
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<tr>
<td>De Leeuw</td>
<td>Telephone vs face to face</td>
<td>n = 14</td>
<td>No overall difference (mean = –0.01)</td>
<td>Year of publication: ‘&lt; 1980’ (mean = –0.03; less bias by face to face), ‘after 1980’ (mean = 0.00)</td>
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<tr>
<td>Years: 1947–1990</td>
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<tr>
<td>Effect estimate: mean weighted product moment correlation</td>
<td>2. Mail q vs face to face</td>
<td>n = 13</td>
<td>Less bias by mail (mean = +0.09)</td>
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<tr>
<td>Studies: n = 52^a</td>
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<tr>
<td>3. Mail q vs telephone</td>
<td></td>
<td>n = 5</td>
<td>Less bias by mail (mean = +0.06)</td>
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<tr>
<td>Richman et al.</td>
<td>Computer vs PAPQ</td>
<td>n = 581</td>
<td>No overall difference (ES = 0.05)</td>
<td>Year of publication: ‘early:1975’ (ES = –0.74; recent: 1996 (ES = –0.08)</td>
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<tr>
<td>Years: 1967–1997</td>
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<tr>
<td>Effect estimate: ES</td>
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<tr>
<td>Studies: n = 61</td>
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<tr>
<td>1. Direct measure of bias</td>
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<tr>
<td>2. Computer vs face-to-face</td>
<td>(Studies – BS: n = 11; WS: n = 17)</td>
<td>n = 92</td>
<td>Less bias by computer (ES = –0.19)</td>
<td>Year of publication: ‘early: 1975’ (ES = 0.12); ‘not alone’ (ES = 0.65)</td>
</tr>
<tr>
<td>Dwight and Feigelson</td>
<td>Computer vs paper and pencil or face to face</td>
<td>IM: n = 45; SDE: n = 32</td>
<td>Less IM bias by computer (ES = –0.08)</td>
<td>Overall ESs for IM bias reduce over time (r = 0.44)</td>
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<tr>
<td>Years: 1969–1997</td>
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<tr>
<td>Effect estimate: ES</td>
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<tr>
<td>Studies: n = 30</td>
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<tr>
<td>2. Computer vs paper and pencil</td>
<td>(studies – BS: n = 33; WS: n = 30)</td>
<td>IM: n = 39; SDE: n = 6</td>
<td>Less IM bias by computer (ES = –0.08)</td>
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<tr>
<td>3. Computer vs face to face</td>
<td></td>
<td>IM: n = 25; SDE: n = 7</td>
<td>No difference in SDE bias</td>
<td></td>
</tr>
</tbody>
</table>

**Commentary:** The effect measure for attitudinal variables compares any one mode with the weighted mean of all responses (not a direct mode vs mode comparison). Differences in size of RE indicate that one mode has more/less bias than another, but not how much. Individual sample size not accounted for in analysis and may have created spurious results.

BS, between subjects; IM, impression management; Mail q, mail questionnaire; PAPQ, paper-and-pencil questionnaire; RE, response effect; SD, social desirability; SDE, self-deception enhancement; WS, within subjects.

^a Includes studies not contributing to social desirability analysis.
However, coding performance (inter-rater reliability) was reported only by de Leeuw and by Richman et al. Difficulties in coding variables with their frameworks was noted by Sudman and Bradburn and by Richman et al., but is probably a ubiquitous problem. The intended coverage of the reviews varied where stated, but is probably generally reflected in the total number of included studies. The Richman et al. review is notable for its attempt to test explicit a priori hypotheses, its operational definition of ‘sensitivity’ and its focus upon features rather than overarching modes. These reviews provide support for the importance of self-administration and consequently impersonality. Richman et al. concluded that there was no overall difference between computer and paper-and-pencil scales. This is consistent with Tourangeau et al.’s model, which directly links computerisation to legitimacy and cognitive burden but not to impersonality. From the first two reviews it is clear that other factors may significantly modify the relationship between mode and social desirability bias. For example, Whitener and Klein found a significant interaction between social environment (individual vs group) and mode of administration (computer:unrestricted scanning vs computer:restricted scanning vs paper-and-pencil).

Acquiescence
Asking respondents to agree or disagree with attitudinal statements may be associated with acquiescence – respondents agreeing with items regardless of there content. Acquiescence may result from respondents taking shortcuts in the response process and paying only superficial attention to interview cues. Knowles and Condon categorise meta-theoretical approaches to acquiescence as addressing either motivational or cognitive aspects of the response process. Krosnick suggested that acquiescence may be explained by the notion of satisficing due to either cognitive or motivational factors. Thus, the role of mode features in varying impersonality and cognitive burden as described above would seem equally applicable here.

There is mixed evidence for a mode feature effect for acquiescence. De Leeuw reported no difference in acquiescence between postal, face-to-face and telephone interviews. However, Jordan et al. found greater acquiescence bias in telephone interviews than in face-to-face interviews. Holbrook et al. also found greater acquiescence among telephone respondents than among face-to-face respondents in two separate surveys.

**What are the consequences of mode feature effects for response quality?**
Several mode feature effects on response quality are listed in Figure 1 and include number of missing data. Computerisation and using an interviewer will decrease the number of missing data due to unintentional skipping. Intentional skipping may also occur and be affected by both the impersonality afforded the respondent and the legitimacy of the survey approach. The model of Tourangeau et al. describes how the reliability of self-reported data may be affected by the cognitive burden placed upon the respondent. De Leeuw provides a good illustration of how the internal consistency (psychometric reliability) of summary scales may be varied by mode features through (1) differences in interview pace and (2) the opportunity for respondents to relate their responses to scale items to each other. The reliability of both multiple- and single-item measures across surveys (and across waves of data collection) may also be affected by any mode feature effects resulting from the psychological appraisal and response processes described above.

Tourangeau et al. highlight how accuracy (validity) of the data may be affected by impersonality and legitimacy. Both unreliable and inaccurate reporting will be represented by variations in the level of an attribute being reported. For example, a consequence of socially desirable responding will be under- or over-reporting of attitudes and behaviour. This may vary depending upon the degree of impersonality and perceived legitimacy.
Additional antecedent features

Two further sets of variables are considered in the framework presented in Figure 1: ‘measurement construct’ and ‘respondent characteristics’. Both represent antecedent features that may further interact with the response process described. For the purposes of this chapter they will be described particularly in relation to health research.

Measurement construct

Objective/subjective constructs

Constructs being measured will vary according to whether they are subjective or objectively verifiable. HRQoL and health status are increasingly assessed using standardised self-report measures [increasingly referred to as patient-reported outcome measures (PROMs) in the health domain]. Although the construct being assessed by such measures may in some cases be externally verified (e.g. observation of physical function), for other constructs (e.g. pain) this may not be possible. Furthermore, the subjective perspective of the individual may be an intrinsic component of the construct being measured. Cote and Buckley reviewed 64 construct validation studies from a range of disciplines (marketing, psychology/sociology, other business, education) and found that 40% of observed variance in attitudes (subjective variable) was due to method (i.e. the influence of measurement instrument) compared with 30% being due to the trait itself. For more objective constructs, variance due to method was lower indicating the particular challenge for assessing subjective constructs.

Sensitivity

Certain clinical topics are more likely to induce social desirability response bias, potentially accentuating mode feature effects. Such topics include sensitive clinical conditions (e.g. human immunodeficiency virus status) and health-related behaviours (e.g. smoking). An illustrative example is provide by Ghanem et al. who found more frequent self-reports of sensitive sexual behaviours (e.g. number of sexual partners in preceding month) using ACASI than with face-to-face interview among attendees of a public sexually transmitted diseases clinic.

Respondent characteristics

Respondent role

In much of the research contributing to the meta-analyses of mode effects on social desirability, the outcome of the assessment was not personally important for study subjects (e.g. participants being undergraduate students). Further methodological research in applied rather than laboratory settings will help determine whether or not mode feature effects are generalisable to wider populations of respondents. It is possible that the motivations of patients (e.g. perceived personal gain and perceived benefits) will reflect their clinical circumstances, as well as other personality characteristics. It is therefore worth investigating whether or not self-perceived clinical need, for example, may be a more potent driver of biased responding than social desirability, and whether or not this modifies mode feature effects.

In a review of satisfaction with health care, the location of data collection was found to moderate the level of satisfaction reported, with on-site surveys generating less critical responses. Crow et al. noted how the likelihood of providing socially desirable responses was commonly linked by authors to the degree of impersonality afforded when collecting data either on- or off-site.

Another role consideration involves the relationship between respondent and researcher. The relationship between patient and health-care professional may be more influential than that between social survey respondent and researcher. A survey request may be viewed as particularly legitimate in the former case and less so in the latter. Response bias due to satisficing may be less of a problem in such clinical populations than in non-clinical populations. Systematic
evaluation of the consequence of respondent role in modifying mode feature effects warrants further research.

**Respondent sociodemographics**

There is some indication of differential mode feature effects across demographic characteristics. For example, Hewitt\(^65\) reports variation in sexual activity reporting between modes [audio-computer-assisted self-administered interview (CASI) and personal interview] by age, ethnicity, educational attainment and income. The epidemiology of different clinical conditions will be reflected by patient populations that have certain characteristics, for example being older. This may have consequences for cognitive burden or perceptions of legitimacy in particular health studies.

**Particular issues in health research**

In considering modes and mode feature effect, we will focus on three issues that may be of particular relevance to those collecting data in a health context: antecedent features, constraints in choice of mode and the use to which the data are being put.

**Particular antecedent features**

Certain antecedent conditions and aspects of the construct being measured may be particularly relevant in health-related studies. Consider the example of QoL assessment in clinical trials of palliative care patients from the perspective of response optimising. Motivation to respond may be high, but may be compromised by an advanced state of illness. Using a skilled interviewer may increase the likelihood of optimising over an approach offering no such opportunity to motivate and assist the patient. Physical ability to respond (e.g. verbally or via a keyboard) may be substantially impaired. This may affect response completeness, but if the overall response burden (including cognitive burden) is increased it may also lead to satisficing. In practice, choice of data collection method will be driven as much by ethical considerations about what is acceptable for vulnerable respondents.

**Are there features of self-reported data collection in health that are particularly different from other settings of relevance to mode feature effects?**

Surveys will be applied in health research in a wide variety of ways, and some will be indistinguishable in method from some social surveys (e.g. epidemiological sample surveys). Some contexts for data collection in health research may be very different from elsewhere. Data collection in RCTs of therapeutic interventions may often include PROMs to assess differences in outcome. How antecedent features in the trial – in particular those associated with respondent role – may influence psychological appraisal and response is hypothesised in Table 2. These antecedent characteristics may potentially either promote or reduce the adverse impact of mode feature effects. The extent to which these effects may be present will need further research, and, at least, would require consideration in trial design.

**Particular constraints on choice of mode**

As in social surveys, mode feature effects will be one of several design considerations when collecting health survey data. Surveying patients introduces ethical and logistical considerations, which, in turn, may determine or limit the choice of data collection method. Quality criteria such as appropriateness and acceptability may be important design drivers.\(^68\) For example, Dale and Hagen\(^67\) reviewed nine studies comparing PDAs with pen-and-paper methods and found higher levels of compliance and patient preference with PDAs. Electronic forms of data collection may offer advantages in terms of speed of completion, decreasing patient burden and enhancing acceptability.\(^68,69\) The appropriateness of different data collection modes may vary by patient
group – for example, with impaired response ability due to sensory loss.\textsuperscript{70} Health researchers need to balance a consideration of mode feature effects with other possible mode constraints when making decisions about data collection methods.

**Particular uses of data**

Evaluating mode feature effects will be particularly important as survey instruments start to play a bigger role in the provision of clinical care, rather than solely in research. PROMs are increasingly being applied and evaluated in routine clinical practice.\textsuperscript{71–73} Benefits have been found in improving process of care, but there is less consistent evidence for impact on health status.\textsuperscript{71,74–76}

Perceived benefits of using such patient-reported outcomes include assessing the impact on patients of health-care interventions, guiding resource allocation and enhancing clinical governance.\textsuperscript{72} Computerised data collection may be especially important if results are to inform actual consultations, but would require suitably supported technology to permit this.\textsuperscript{77,78} With only mixed evidence of clinical benefit, Guyatt \textit{et al.}\textsuperscript{76} highlight computerised-based methods of collecting subjective data in clinical practice as a lower-cost approach.

In this clinical service context, psychological responses such as social desirability bias may vary according to whether patient data are being collected to inform treatment decision-making or clinical audit. Method of data collection may similarly play a role in varying the quality of response provided. However, routinely using subjective outcome measures in clinical practice will require a clear theoretical basis for their use and implementation, and may necessitate additional training and support for health professionals and investment in the technology to support its effective implementation, which is, preferably, cost neutral.\textsuperscript{79–82} Overall, though, it may be that any biasing effect of mode feature may be less salient in situations where information is being used as part of a consultation to guide management, and may be more so where data are being collected routinely across organisational boundaries as part of clinical audit or governance.
Managing mode feature effects in health

Managing mode feature effects requires identification of their potential impact. This chapter has focused upon response quality as one source of error in data collection. Two other sources of error influenced by mode are ‘coverage error’ and ‘non-response error’. In the former, bias may be introduced if some members of the target population are effectively excluded by features of the chosen mode of data collection. For example, epidemiological surveys using random digit dialling, which exclude people without landline telephones, may result in biased estimates as households shift to wireless-only telephones. Response rates vary by mode of data collection and different population subgroups vary in the likelihood of responding to different modes. For example, Chittleborough et al. found differences by education, employment status and occupation among those responding to telephone and face-to-face health surveys in Australia.

Social surveys commonly blend different modes of data collection to reduce cost (e.g. by a graduated approach moving from cheaper to more expensive methods). Mixing modes can also maximise response rates by, for example, allowing respondents a choice about how they respond.

In the long term it may prove possible to correct statistically for mode feature effects if consistent patterns emerge from meta-analyses of empirical studies. Alternatively, approaches to reducing socially desirable responding have targeted both the question threat and confidentiality. An example of the latter is the randomised response technique, which guarantees privacy. Another approach is the use of goal priming (i.e. the manipulation and activation of individuals’ own goals to subsequently motivate their behaviour), where respondents are influenced subconsciously to respond more honestly.

Evaluating and reporting mode feature effects

As described above, the evaluation of data collection method within individual studies is usually complicated by the package of features representing any one mode. Groves et al. described two broad approaches to the evaluation of effects due to mode features. The first and more pragmatic strategy involves assessing a package of features between two or more modes. Such a strategy may not provide a clear explanation for resulting response differences, but may satisfy concerns about whether or not one broad modal approach may be replaced by another. The second approach attempts to determine the features underlying differences found between two modes. This theoretically driven strategy may become increasingly important as data collection methods continue to evolve and increase in complexity.

As global descriptions of data collection method can obscure underlying mode features, comparative studies should describe these features more fully. This would enable data synthesis, providing greater transparency of method and aid replication.

Summary

This chapter has considered how features of data collection mode may impact upon response quality, and key messages are summarised in Box 1. It has added to a model proposed by Tourangeau et al. by drawing apart psychological appraisal and response processes in mediating the effect of mode features. It has also considered other antecedent features that might influence response quality. Mode feature response effects have been most thoroughly reviewed empirically in relation to social desirability bias. Overall effects have been small, although evidence of significant effect modifiers emphasises the need to evaluate mode features rather than simply overall mode. A consistent finding across the reviews is the important moderating effect of year of publication for comparisons involving both telephone and computers. Therefore, mode feature comparisons are likely to remain important as new technologies emerge for collecting data. Although much of the empirical research underpinning the reviewed model has been generated...
within other academic domains, the messages are nonetheless generally applicable to clinical and health research.

Future evidence syntheses may confirm or amend the proposed model, but this requires as a precursor greater attention to theoretically driven data collection about mode features. The current theoretical review framework, therefore, provides the basic analytic structure for the analysis and a basis upon which emergent results may be interpreted (see Box 1). In particular, the emphasis upon mode features is a key contributor to this analytic model.
Chapter 3

Review methods

The methods used to evaluate the impact of features of mode of data collection on subjective outcome measures follow that of a systematic review. Owing to the large body of literature relating to survey methodology which is published outside the health research arena, all studies that incorporate a mode comparison will be included, regardless of setting. This leads to a broad search strategy covering a wide range of disciplines. In order to target methodological studies, some innovations in search strategy have been undertaken that separate out the process from the traditional reviews of the effectiveness of interventions.

Identifying the literature

Inclusion/exclusion criteria

The inclusion/exclusion criteria for a study to be included in the review were as follows:

- There is evidence of a comparison between two modes of data collection of either the same question, or set of questions, referring to the same theoretical construct.
- The construct compared is subjective and cannot be externally validated.
- The analysis of the study contains an explicit reference to a comparison.
- Data collection is quantitative, i.e. uses structured questions and answers.

This can include studies in which mode comparisons were made, even if not the main purpose of the study.

Studies were excluded from the review if they involved:

- a comparison between a quantitative measure and one or more qualitative data collection methods/analyses (e.g. unstructured interviews, focus groups)
- a comparator derived from routine clinical records – unless explicit reference to specific self-reported construct is made within those records
- a comparison between the response of two different judges, i.e. comparing a response from an individual with that made by someone other than the responder, for example a clinician providing a diagnosis.

Subjective measures are defined as those in which the perspective of the individual is an intrinsic component of the construct being measured. Comparisons between two different perspectives (even on the same construct) are therefore excluded.

Year of publication

All databases were searched from the earliest point in time until the end of 2004. This was based on the last complete year available to the researchers at the point at which the search was undertaken.

Language and location

No studies were excluded owing to language or country of origin to allow inclusion of as much innovation in design and novel mode application as possible. It is known that the perceived ‘gold
standard’ method of data collection will vary, especially in relation to approaching respondents in their homes.99 In some cultures, a face-to-face interview is perceived as more acceptable than calling on the telephone.90 In addition, matters of privacy and over-use of mass marketing schemes have changed the availability of telephone numbers and ability to contact. For example, the use of automated marketing technology in the UK has given rise to ‘preference’ services offered by telecommunications companies and the Post Office, whereby registered marketing companies cannot gain access to the recipient.

This chapter will document the development, piloting and optimisation of the search strategy, the three-phase selection process and the methods of synthesis for the data extracted.

Databases
Owing to the broad range of disciplines outside the health sector literature which cover survey methodology, a subject-free approach was required to collate evidence from all research. However, databases were chosen based on subject area to allow as comprehensive a selection as possible. A full list of databases used in the review can be found in Appendix 1 (see Table 21).

MEDLINE was used for the initial development and optimisation of the search strategy. It was decided that grey literature and grey databases would not be searched, as the effort required to retrieve such information usually outweighs the gains.91 Therefore, only journal articles and conference abstracts cited within journal supplements were included in the review process.

Search strategy
Guides for the development and creation of search strategies used in systematic reviews in defined areas have been well described.92 However, guides do not exist for searching such a diffuse and multidisciplinary literature base. Therefore, the search strategy for the present review was continually optimised using an iterative process. Initially, an extensive development phase was carried out, followed by the main search and retrieval phase.

From previous literature reviews in the area of survey research2,93 it was shown to be possible to systematically identify a body of literature describing the effects of differences in modes of data collection. However, studies that use only a single mode of data collection are not of interest and, therefore, in order to focus the search strategy, a matrix approach was developed. The matrix was intended to facilitate the search for articles that had two or more modes of data collection. Each column and row of the matrix consisted of a collection of terms relating to a single mode (e.g. postal, survey, mail). Only the off-diagonal terms were considered for inclusion (highlighted cells in Table 3). This used Boolean terminology: Group 1 AND (Group 2 OR 3 OR 4 OR 5 OR 6 OR 7

<table>
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<tr>
<th>TABLE 3</th>
<th>Illustration of matrix approach to identification</th>
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<td>Mode of data collection</td>
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<td>Mode of data collection</td>
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OR 8 OR 9 OR 10). For example, this would identify any paper that had terms relating to desktop computer use and any one of face to face, paper and pencil, etc.

Initially, 10 different types of data collection mode were identified, which were defined as 'data collection groups'. A list of search terms was generated for each group. From these categorisations, one row and column (representing paper-and-pencil administration) was selected and all abstracts identified (759) were screened and the terms and categorisations tested to see if a more specific search strategy would have identified the same studies. On the basis of this, the data collection groups were revised from 10 to 8 as follows:

1. technology assisted (computer and PDA combined)
2. internet based
3. antonym of technology
4. paper-and-pencil administration (combined with mail)
5. fax administration
6. telephone administration
7. in-person administration
8. unspecified mode.

It became apparent that there was an ordered use of language in all articles, allowing a grammatical framework to be applied to the search terms within the data collection groups. Search terms relating to different modes of data collection could be described as a nominal phrase, consisting of a compound noun and one or more compound adjectives. New modes of data collection have evolved with the creation of new technologies, and, instead of developing new nouns, existing nouns have been modified by the development of compound nouns, qualified by compound adjectives, for example computer-assisted telephone interview (CATI).

Search terms generated in the initial searches were allocated to the different data collection groups by linking the compound adjective to the group with which it was most associated. The final search terms for each data collection group are in Appendix 1.

Medical subject headings (MeSH) were utilised where available. The specific thesaurus terms used in each database and field codes used to implement the matrix section of the search strategy are detailed in Appendix 1 (see Table 21), concerning health and evidence-based medicine, social sciences, and economics and other, respectively. The use of MeSH can seriously influence the noise in the search strategy (the number, and type of citations retrieved) due to the branching hierarchical classification. For example, when locating articles related to methodological issues the search term 'method' is prolific in the introduction, method, results and discussion (IMRaD)-constructed abstracts, whereas the more specific term 'methodolog$' searched in the title, abstract and keywords of the article yielded more precise results.

The strategy was implemented in MEDLINE from the beginning of 1966 to the end of 2004, and all articles were subsequently screened for relevance. The screening accompanied an iterative process identifying new research-specific terms. The iterative process generated 24 new nominal phrases that were added to the appropriate groups, and one new group was identified pertaining to the use of video. No clear distinction was developed between online and offline computerised methods, therefore the terms in the internet-based group were merged with the technology-assisted data collection methods. The strategy was then re-implemented to screen for new, previously unidentified articles.

In order to focus the search on studies that were comparisons of modes, rather than just studies that happened to report two modes, the studies identified from the searches above were limited to those that used terms suggestive of a comparison (e.g. comparison, versus, trial, evaluation)
and general terms relating to data collection (e.g. administration, survey, assessment). Therefore, only studies that combined all three domains were included for further consideration (Figure 2).

Following the successful development of the search strategy within MEDLINE, the same strategy was implemented within all the specified databases, allowing for changes in field codes and thesaurus terms as described in Appendix 1.

Citation information and abstracts were downloaded from the selected databases and imported into an EndNote (Thomson Reuters, CA, USA) database. At each stage of the download, the number of articles requested for download and the numbers of articles actually downloaded were checked for consistency. Duplicate citations were removed using the EndNote Version 9 'Find Duplicate' function. Citations were considered duplicates if either:

- the title field exactly matched another citation, or
- the author, year, journal, volume, issue and page numbers exactly matched.

**Review process**

The abstracts (and titles only for some foreign-language papers with no English abstract) were reviewed against the inclusion/exclusion criteria. No assessment was made at this stage as to the subjectivity of the measures presented. Full papers were retrieved for all selected abstracts and then screened again relating to more detailed inclusion criteria relating to the measures used. Papers that were still included were reviewed in full and detailed data extracted (Figure 3). The datasheets used for full-paper screening and data extraction are given in Appendix 2. The screening and data extraction stages were combined for foreign-language papers.

At each stage, abstracts or papers were reviewed by a single reviewer after a period of training (Figure 4). Training for each stage included an assessment of reliability and sensitivity. Training and testing sets of abstract/papers were used. This was repeated for hits from different databases to allow for reassessment with different types of study and abstract layout.

**FIGURE 2** Conceptualisation of search strategy.
Rigorous training ensured high reliability of the screening process. To quantify this, the efficacy of training was assessed by calculating the area under the curve (AUC) devised from the receiver operating characteristic curve (ROC). The AUC was calculated against a 'gold standard' of exact matches arrived at by consensus. Having a sensitive process was considered more appropriate than overall agreement, with a focus on over-including (where in doubt in the early stages) being important to avoid missing key studies.

Three reviewers undertook abstract screening (AS, GG and KH). After the triplicate screening of 750 abstracts (three sets of 250) from MEDLINE, the ROCs were calculated, generating AUC scores: AS = 0.865, GG = 0.954, KH = 0.970. Training was repeated for PsycINFO, and 750 triplicate-screened abstracts generated AUC scores: AS = 0.88, GG = 0.92, KH = 0.90. Five reviewers undertook the initial screening of the full papers (AS, GG, KH, MR and CS). Training was carried out with 20 articles and reviewed independently. Consensus was achieved through discussion of included and excluded studies. Then a subsequent set of 20 studies were reviewed independently and the sensitivity of all reviewers was 100%. Data extraction and quality assessment were undertaken by three reviewers (GG, NC and RI). Training was carried out on two sets of 20 papers, giving AUC scores of GG = 0.823, NC = 0.802 and RI = 0.790.
Data extraction

The final extraction stage was carried out using a series of forms (see Appendix 2). These forms were circulated to all members of the study management team for comment and changes were implemented accordingly. As with each stage of the reviewing process, a training phase was completed. The data extraction was comprehensive because of the wide-ranging and diverse nature of the articles selected. Items for data extraction were selected to be as inclusive as possible; the details of each included study were captured under the following headings:

1. population and design (data forms 2 and 3)
2. mode description (data form 4)
3. measure description (data form 5)
4. comparison (data form 6).

Every paper reviewed had one form describing the setting and design of the study and its overall quality. For the other data forms, variable numbers were completed depending on the number of modes and measures compared. These were then linked using the unique study ID number.

Modes were put into a general categorisation, as well as classified by their mode features. The mode features were based on the theoretical framework developed in Chapter 2 and additional features indicated as possibly related to response differences in the literature. The four main features from the theoretical framework were:

- administration (self or interviewer)
- telephone contact
- computerisation
- sensory stimuli (auditory, visual or both).

The first mode feature of administration is relatively self-explanatory. Modes in which an interviewer was recorded and then either played down the telephone, on video or on a computer are still classified as self-administered, as the control of the interview is with the respondents; for example, they can pause and play or stop at will.

The use of telephone could be by an interviewer or via an automated dial-up service for administration. The use of a computer can be in the form of a CATI, a CAPI or computer-based self-administration, such as a disk by post or a web survey. Sensory stimuli are coded on the basis of having purely auditory stimuli, such as simple telephone and face-to-face interviews; purely visual stimuli such as paper-based questionnaires or simple web surveys; or modes that combine both, such as face-to-face interviews with use of prompts such as flash cards or web-based surveys with a video/audio component.

Other features were coded to be tested for inclusion in the model. These related to the perceived legitimacy, such as how the measure was delivered to the respondent. This could be by telephone, in person or via the post/e-mail/web. Although the majority of telephone and face-to-face administered modes would have the same delivery as administration, for some studies these will be different, for example more laboratory-based studies in which all modes are introduced in person, but may still be completed as self-complete questionnaires or on a computer.

A number of other factors related to perceived anonymity, such as the mode of response provided, whether or not others were present during completion and whether or not anonymity was specifically protected. The ability to backtrack was also collected as a possible contributing factor to the level of cognitive burden.
For statistical data extraction, where standard deviation (SD) data were not presented, they were imputed from $p$-values, confidence intervals (CIs) or test statistics where available. Where information about scales, such as number of items, scoring, etc., was not provided in papers, the original source references for those studies were accessed for information.

**Quality assessment**

In order to assess the quality of the evidence contributing to this review, each paper was assessed for methodological quality. Assessing the quality of evidence becomes particularly challenging in the reviews of studies having diverse methodologies. In this particular review, RCTs were not necessarily expected, and so a more generic quality assessment tool was needed. Two tools were identified,\(^94,95\) which provided quality checklists for studies other than RCTs.

Downs and Black\(^94\) created a checklist for both randomised and non-randomised studies, focusing on health-care interventions. The checklist consisted of 27 items from five subscales:

1. **Reporting**  Do the findings allow the reviewer to draw unbiased conclusions?
2. **External validity**  Can the findings be generalised?
3. **Bias**  Have potential biases been addressed and mentioned?
4. **Confounding**  Have possible confounders been addressed and reported?
5. **Power**  Could the findings be due to chance?

The tool, scored on a dichotomous scale, has good face validity, demonstrates inter-rater reliability and correlates well with an existing validated checklist, the Quality Index.\(^96\) The checklist provides a detailed profile of both randomised and non-randomised studies.

Kmet \(^{95}\) et al. took this process one step further by developing tools for both quantitative and qualitative research. The process, scored on a scale of zero to 2, evaluated the methodological choices and the clarity of reporting in relation to potential biases. However, the authors tested the checklist on only 10 articles, allowing a limited inter-rater reliability analysis.

The current tool was based upon the previous two checklists, with some modifications. The checklist of Downs and Black\(^94\) is detailed containing 27 items, but is heavily weighted towards randomised designs. The Kmet \(^{95}\) et al.\(^95\) checklist, although shorter at 14 items, focuses on intervention studies, which was not appropriate for this review. Therefore, it was necessary to create a checklist designed specifically for this present review that was more appropriate to both the methodological nature of the topic and the diverse literature base. The resulting checklist (see Appendix 2, datasheet 7) contained 18 items scored on three levels, yes (2), partial (1) and no (0), with three questions containing 'non-applicable' categories for specific study designs. Scores are summated across each item providing a percentage score, allowing consideration for the non-applicable items.

The piloting of quality assessment allowed testing of the inter-rater reliability. Both main reviewers (GG and RI) separately scored the quality of 20 papers included in the full data extraction phase. The scoring of each paper was carried out after the main descriptive and quantitative extraction of data from the papers. The detailed reading required for the data extraction process facilitated judgements of quality. As such, the checklist was quick and easy to complete, taking approximately 2 minutes per paper. Agreement between GG and RI was good, with $\kappa$-values on individual items ranging from 0.61 to 0.85. A paired-sample $t$-test on total scores demonstrated no significant differences between the reviewers (mean difference = 1.17, SD = 4.50, $p = 0.8$).
**Publication bias**

The conceptual framework for publication bias being based on journals and investigators not wanting to publish ‘negative’ studies is unlikely to apply in the case of mode comparison studies. The consideration that two modes are the same or different is equally likely to be newsworthy. Therefore, it is more likely that gaps in publications are likely to appear due to methodological reasons rather than outcome (poorly designed studies) or sample size (too small studies).

**Evidence synthesis**

An overview of the studies identified will be presented descriptively highlighting the different mode features identified in the theory review. Those with appropriate data will be subjected to quantitative methods of synthesis using exploratory metaregression to identify the association between mode features and differences in response.

The primary analysis based on three key summary statistics is calculated for each comparison. These are:

- the absolute difference between the means (standardised) of the two modes
- the ratio of the largest to the smallest variance of the two modes
- the ES (absolute mean difference/SD) between two modes.

This allows for separate consideration of the accuracy and precision of the measures collected by the two modes as well as the more usual ES which combines both. For the first analysis, the mean differences need to be standardised to allow for measures on different scales to be combined. Using the highest and lowest possible scores on each scale, these were standardised to a 0–100 scale.

\[
\text{Standardised score} = \frac{(\text{actual score} - \text{minimum value})}{(\text{maximum value} - \text{minimum value})} \quad [\text{Equation 1}]\]

Where the average scores per item are used in summary statistics, the minimum and maximum values per item were used to standardise. The absolute value of the difference is used as, when combining many different outcomes, the direction of difference is meaningless.

For the second analysis the ratio of the two variances is already on a standardised scale as the largest variance is being presented as a proportion of the smallest. Similarly, the ES is a standardised statistic with the absolute mean difference expressed as a proportion of the SD. The pooled SD from the two modes was used in the calculation of the ES.

Between- and within-subject studies were analysed together, controlling for the study design. Analysis was conducted at two levels to account for clustering of comparisons within a study. This allowed for study-level characteristics, measures characteristics and mode features to be considered together. The modelling approach assessed the four main mode features from the theoretical review, then tested the addition of other candidate features and then assessed model fit including other possible moderators of effect and identified interaction. Studies were categorised whether or not they were designed to show a difference on a mode feature. For example, this meant that a web versus a postal survey would have been coded as no difference on the feature of administration, whereas a web survey versus a telephone interview would have been coded as showing a difference. These differences were then used as explanatory variables in the models.

Sensitivity analysis explored the impact of weighting by quality scores (rather than using as an explanatory variable), as well as weighting by functions of the sample size and the pooled SD.
Statistical methods for individual within-group comparison studies for two methods of measuring the same entity have been debated extensively. This is particularly so in the field of clinical measurement where two clinical tools (e.g. thermometers) are compared on the same patients.\textsuperscript{99–101} These techniques have varied from relatively simple methods for assessing accuracy and precision of instruments (e.g. limits of agreement and Bland–Altman plots\textsuperscript{100}) to more complex modelling (e.g. structural equation modelling). Williamson \textit{et al.}\textsuperscript{102} developed two approaches to estimating combined limits of agreement\textsuperscript{102} and the Mantel–Haenszel approach is presented for the two most frequently occurring scales, the Short Form questionnaire-36 items (SF-36) and MMPI. Studies that are between-group comparisons of these two sets of measures are also subjected to a standard random-effects meta-analysis. The original proposal was to undertake a review of the differences between studies of a single mode using SF-36; however, this was replaced with the meta-analysis above as being more appropriate given the number of studies identified which directly compared two modes using the SF-36. The MMPI was added owing to the number of studies reporting this outcome.

Analysis was undertaken using SPSS 14.0 (SPSS Inc., Chicago, IL, USA), MLwiN 1.1 (Centre for Multilevel Modelling, University of Bristol, Bristol, UK) and RevMan 5 (The Cochrane Collaboration, The Nordic Cochrane Centre, Copenhagen, Denmark).

\textbf{Changes from original proposal}

A number of minor changes were made to the original proposal, the search strategy was developed and refined from that in the original proposal when the theoretical review suggested that it was simplistic to simple categorise by crude mode and the training plan to ensure that individual reviewers was developed to incorporate all stages of review instead of simply the data extraction phase as stated. This was undertaken on a slightly smaller number of papers (20 rather than 25) than originally stated as agreement was good and individuals had already received considerable training in earlier phases. The major change was that the review of single-mode studies for SF-36 was replaced by a more detailed analysis of the mode comparison studies identified for that measure and also the MMPI. This decision was based on the numbers of studies identified.

This study is reported in accordance with reporting standards for systemic reviews and the PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) checklist\textsuperscript{103} is included in Appendix 4.
Chapter 4

Results

The search strategy produced a total of 63,305 citations downloaded from the various databases, of which 39,253 were unique (Figure 5). These articles had their titles and abstracts reviewed, with 2156 articles being selected for retrieval in full. The full articles were then screened prior to detailed data extraction. The process excluded 1559 papers (see Table 4 for details).

Studies excluded from the review

Table 4 shows the number of papers excluded from initial screening of the full 2156 papers and the reasons for their exclusion.

The most common reason for exclusion (44%) was that the paper did not contain a mode comparison. A number of studies (12%) described use of multiple modes of data collection; however, these were for different outcomes often measured at different time points. The next

---

**FIGURE 5** Flow diagram of study identification.

**TABLE 4** Reasons for exclusion from the initial screen of the full paper

<table>
<thead>
<tr>
<th>Reason</th>
<th>No.</th>
</tr>
</thead>
<tbody>
<tr>
<td>No mode comparison</td>
<td>691</td>
</tr>
<tr>
<td>Mode comparison, but not comparing the same construct</td>
<td>91</td>
</tr>
<tr>
<td>Comparison of different judges</td>
<td>458</td>
</tr>
<tr>
<td>Measuring or comparing a behavioural construct only</td>
<td>230</td>
</tr>
<tr>
<td>Review (not primary study)</td>
<td>89</td>
</tr>
<tr>
<td>Total number of papers excluded at first stage</td>
<td>1559</td>
</tr>
</tbody>
</table>
most common reason (29%) was that the article referred to a comparison of two different judges, the most common of these being clinical diagnostic interviews for psychiatric disorders. As this incorporation of a second individual’s judgement into one mode could invalidate the comparison, all structured clinical interviews have been excluded. The next largest group (15%) was that of papers that compared a behavioural construct only. These papers focused mainly on sensitive behaviours, such as smoking, sexual behaviour and drug taking. All of these papers were retrieved at abstract stage to be checked for any subjective component being reported, even when the main focus of the study was on measuring behaviour. Papers which solely focused on behaviour were excluded at this stage, whereas those that included some subjective elements were retained (e.g. being scared by your level of drinking would be included but the amount of alcohol drunk would not).

Of the 597 articles for which data extraction was undertaken, a total of 216 were also excluded (Table 5).

The most common reasons were that the construct being compared was not subjective (36%) or that it was judged by two different individuals (36%) (e.g. patient and clinician or parent and child). The next most common was if the paper contained no mode comparison (18%). This commonly occurred in studies in which there were two modes of data collection but no common data collected through multiple modes and therefore no mode comparison. An additional 13 papers (6%) were excluded as they only reported response rates and had no information on the actual responses given.

Thirty foreign-language articles were retrieved in full on the basis of their English title and abstract (where available). These were then screened for inclusion and data extracted where appropriate by one of the main reviewers (GG or RI) and a translator. The languages included Chinese, Danish, Dutch, French, German, Japanese and Spanish. During this process, it was found that 10 papers were to be excluded. Five further papers (two in Slovenian, two Russian and one Czech) were unable to be translated owing to the unavailability of a translator.

**Description of included studies**

Studies from 381 articles met the inclusion criteria for the review. There has been an increase in the number of published mode comparison studies over recent years (Figure 6). This increase in studies may represent many factors directly or indirectly linked to the methodology of mode comparison experiments. The first influence relates directly to the increase in technological options available to the survey researcher. However, direct factors such as the increase in the number of journals, particularly those that are electronic only, have led to a general increase in publications levels.

**TABLE 5** Reasons for exclusion at the data extraction stage

<table>
<thead>
<tr>
<th>Reason</th>
<th>No.</th>
</tr>
</thead>
<tbody>
<tr>
<td>No mode comparison</td>
<td>39</td>
</tr>
<tr>
<td>Comparison of different judges</td>
<td>79</td>
</tr>
<tr>
<td>Measuring or comparing a behavioural construct only</td>
<td>79</td>
</tr>
<tr>
<td>Focuses on response rates only</td>
<td>13</td>
</tr>
<tr>
<td>Review</td>
<td>1</td>
</tr>
<tr>
<td>Unable to translate</td>
<td>5</td>
</tr>
<tr>
<td><strong>Total number of papers excluded</strong></td>
<td><strong>216</strong></td>
</tr>
</tbody>
</table>
Source of publication

Data were collected on the subject area in which the mode comparison was carried out. Most mode comparison studies were published in the area of health (n = 201, 53%). The next largest area of study for mode comparisons was psychology (n = 86, 23%) and social sciences (n = 55, 14%). The rest of the studies were focused on business (n = 16, 4%), statistics (n = 14, 4%) and education (n = 9, 2%).

Country and language of data collection

The review was not restricted by location of study or language, Table 6 shows the distribution of the study locations. A large proportion of the studies were carried out in North America (n = 236, 62%), with 112 (29%) studies being carried out in Europe and 38 (10%) of those were from the UK.

The language of data collection was predominantly English (n = 274, 72%), although this was mostly inferred as it was clearly stated in only 30 (8%) of these papers. The other languages used were predominantly European in origin, with French, German, Dutch and Spanish being the most frequent.

Study design

Studies were categorised based on the incorporated study design, either within subjects or between subjects. In total, 52% of studies (n = 200) were designed to provide a between-group comparison of modes, whereas 47% (n = 180) were within-group comparisons. Studies that were crossover by design have been included in the grouping in which they provided data for comparison (predominantly within groups).

Data were collected on whether the studies randomised either the mode an individual received (between-group studies) or the order in which modes were received (within-group studies). In total, 147 studies (39%) had used randomisation, with a higher proportion of between-group studies (n = 83, 42%) than within-group studies (n = 64, 36%) using this form of allocation. Studies which did not use randomisation used other forms of allocation such as drawing samples.
from separate sampling frameworks (e.g. separate population surveys\textsuperscript{104} in between-group studies and systematic allocation (e.g. alternating\textsuperscript{105,106}) for within-group studies. A relatively large number of within-group studies presented the modes under evaluation in exactly the same order to all participants ($n = 95, 53\%$).\textsuperscript{107,108}

The 381 papers included in the review described 489 different samples. Some studies compared response on samples derived from two different sources (e.g. online survey panel compared with random-digit dialling). The methods for sampling demonstrated a dominance of two distinctly different approaches either by convenience ($n = 155, 32\%$) or targeting a specific group of participants, for example on a clinic list ($n = 257, 53\%$).

### The measurement of study quality

The quality of every study included in the review was assessed utilising an 18-item tool specifically designed for the present review. The tool measures quality of quantitative studies irrespective of study design. Overall scores were generally high (Figure 7). However, certain items

<table>
<thead>
<tr>
<th>Country</th>
<th>No. of studies</th>
</tr>
</thead>
<tbody>
<tr>
<td>USA</td>
<td>201</td>
</tr>
<tr>
<td>Canada</td>
<td>36</td>
</tr>
<tr>
<td>UK</td>
<td>38</td>
</tr>
<tr>
<td>Germany</td>
<td>19</td>
</tr>
<tr>
<td>Australia</td>
<td>13</td>
</tr>
<tr>
<td>Netherlands</td>
<td>11</td>
</tr>
<tr>
<td>France</td>
<td>9</td>
</tr>
<tr>
<td>Sweden</td>
<td>7</td>
</tr>
<tr>
<td>Denmark</td>
<td>6</td>
</tr>
<tr>
<td>Spain</td>
<td>6</td>
</tr>
<tr>
<td>Norway</td>
<td>5</td>
</tr>
<tr>
<td>Switzerland</td>
<td>5</td>
</tr>
<tr>
<td>Belgium</td>
<td>2</td>
</tr>
<tr>
<td>Israel</td>
<td>2</td>
</tr>
<tr>
<td>Turkey</td>
<td>2</td>
</tr>
<tr>
<td>Austria</td>
<td>1</td>
</tr>
<tr>
<td>Brazil</td>
<td>1</td>
</tr>
<tr>
<td>China</td>
<td>1</td>
</tr>
<tr>
<td>Croatia</td>
<td>1</td>
</tr>
<tr>
<td>Finland</td>
<td>1</td>
</tr>
<tr>
<td>Hong Kong</td>
<td>1</td>
</tr>
<tr>
<td>Ireland</td>
<td>1</td>
</tr>
<tr>
<td>Italy</td>
<td>1</td>
</tr>
<tr>
<td>Japan</td>
<td>1</td>
</tr>
<tr>
<td>Mexico</td>
<td>1</td>
</tr>
<tr>
<td>Unknown</td>
<td>14</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td><strong>386</strong></td>
</tr>
</tbody>
</table>

\* Three studies were carried out in two countries and one study in three. Total number of studies = 381.
showed a higher percentage of poor ratings than the others. These were the items relating to clear
descriptions of participants (22% poor), group allocation (50% poor), appropriate consideration
given to the impact of timing of data collection (27% poor) and reporting of variances for
results (35% poor). However other items had extremely high scores such as having a clearly
stated hypothesis (89% good), the study design described and appropriate (83% good) and the
conclusions supported by the results (81% good).

**Measures used**

In total, the 381 papers provided 1282 measure descriptions. Thirty per cent of studies considered
only a single measure, with one study comparing 21 different measures (Figure 8). The term
measure did not relate solely to one tool, but to the subscales within the measure, for example a
study that reported using all subscales of the SF-36 would represent eight measures.

![Figure 7](image-url)  
**FIGURE 7** Frequency distribution of percentage quality scores for included studies.

![Figure 8](image-url)  
**FIGURE 8** The number of studies by number of measures reported.
Each measure described was categorised as whether it concerned a health-related area or not. Measures such as QoL symptoms, as well as those relating to general mental well-being (anxiety, etc.) were classified as health and those measuring societal attitudes, personality and willingness to pay were classified as non-health. Of the 1282 measures described, 733 (57%) were classified as being health related.

To examine further the type of constructs measured, the measures were categorised based upon the psychological construct being measured. Studies measuring personality ($n = 257$, 20%) and specific aspects and dimensions of QoL ($n = 215$, 17%) were the most common. The most frequently occurring scales were the SF-36 (17 studies) and the MMPI (nine studies), which have 8 and 14 subscales, respectively, and therefore dominate the QoL and personality assessment categories. It should also be acknowledged that the categorisation is as driven by the description from the scale developers, and for some scales there may be little difference, for example, between the types of measures which have been classified as QoL and those classified as functional health status.

**Modes evaluated**

In total, the 381 papers described 801 modes. All studies provided a comparison between at least two modes (because of the inclusion criteria); however, some studies compared more, with 35 (9%) comparing three modes and two studies (1%) comparing four modes.

Each mode can be roughly categorised into one of four groups by main delivery method. These can be considered to be:

- computer (including web)
- paper
- telephone
- in person (face to face).

Although the features identified in the theoretical review cut across these categories, all the comparisons identified are between rather than within these categories. The total numbers of papers (and comparisons) by comparison group are given in Table 7.

As well as the relatively simplistic categorisation above, a more detailed level of information was obtained relating to specific features of the survey mode. This stratification was defined by the work of Tourangeau *et al.*[^8] and discussed in Chapter 2. This theoretical framework defines four

<table>
<thead>
<tr>
<th>Comparison</th>
<th>No. of studies</th>
<th>No. of comparisons</th>
<th>Comparisons per study: mean (median)</th>
<th>Range</th>
</tr>
</thead>
<tbody>
<tr>
<td>Computer vs paper</td>
<td>161</td>
<td>665</td>
<td>4.1 (2)</td>
<td>1 to 23</td>
</tr>
<tr>
<td>Computer vs telephone</td>
<td>12</td>
<td>17</td>
<td>1.4 (1)</td>
<td>1 to 3</td>
</tr>
<tr>
<td>Computer vs person</td>
<td>22</td>
<td>50</td>
<td>2.3 (2)</td>
<td>1 to 11</td>
</tr>
<tr>
<td>Paper vs telephone</td>
<td>74</td>
<td>290</td>
<td>3.8 (2)</td>
<td>1 to 36</td>
</tr>
<tr>
<td>Paper vs person</td>
<td>106</td>
<td>367</td>
<td>3.5 (2)</td>
<td>1 to 24</td>
</tr>
<tr>
<td>Telephone vs person</td>
<td>52</td>
<td>143</td>
<td>2.8 (1.5)</td>
<td>1 to 11</td>
</tr>
<tr>
<td>Overall</td>
<td>383[^a]</td>
<td>1522</td>
<td>4.0 (2)</td>
<td>1 to 36</td>
</tr>
</tbody>
</table>

---

[^8]: Tourangeau *et al.*

[^a]: Some papers appear in more than one category.
main mode features (administration, telephone contact, computerisation and sensory stimuli). Additional mode features related to other potential mediating factors are also explored whether or not they explain variation over and above that explained by the four main features.

### Four main mode features

Of the total number of comparisons made, 667 (44%) involved a comparison between administration by an interviewer and self-completion. Telephone contact was one of the differences between modes for 440 (29%) of the comparisons (Table 8). Computers were incorporated in data collection of one mode in 803 (53%) comparisons. There was a difference in the main sensory stimuli in 714 (47%) comparisons.

### Other possible mode features

Other features that were considered were the methods of delivery and response for the measure, whether or not the measure was completed ‘online’ (i.e. inputted through a technological device, which is connected to another technological device in ‘real time’, such as a telephone connected to another telephone or computer), who was physically present during completion (interviewer/other), the degree of anonymity of the process and the ability to backtrack through questions, and whether the response was oral, written or by means of electronics (e.g. pushing buttons).

The presence of others (not including the interviewer/researcher) and the ability to backtrack through a questionnaire were only explicitly mentioned in 6% of comparisons. Although reported in more studies, the degree of anonymity was different in only 13 comparisons (1%). None of these three features is therefore included in further modelling.

### Possible mediators

The key theoretical mediating factors within the model presented in Chapter 2 are impersonality, legitimacy and cognitive burden. There were no direct measures of impersonality that are reported in the studies, and any indirect assessment is instead inferred from the description of the mode features above. The issue is similar for legitimacy, although information on the source of approach for a study was recorded in 350 papers (92%). However, the source was a public body

<table>
<thead>
<tr>
<th>Mode feature</th>
<th>Difference</th>
<th>No difference</th>
<th>Missing</th>
</tr>
</thead>
<tbody>
<tr>
<td>Administration</td>
<td>667</td>
<td>855</td>
<td>0</td>
</tr>
<tr>
<td>Telephone</td>
<td>440</td>
<td>1082</td>
<td>0</td>
</tr>
<tr>
<td>Computer</td>
<td>803</td>
<td>719</td>
<td>0</td>
</tr>
<tr>
<td>Sensory stimuli</td>
<td>714</td>
<td>808</td>
<td>0</td>
</tr>
<tr>
<td>Delivery method</td>
<td>686</td>
<td>836</td>
<td>0</td>
</tr>
<tr>
<td>Presence of interviewer/researcher</td>
<td>672</td>
<td>850</td>
<td>0</td>
</tr>
<tr>
<td>Online/offline</td>
<td>523</td>
<td>999</td>
<td>0</td>
</tr>
<tr>
<td>Response method</td>
<td>1386</td>
<td>136</td>
<td>0</td>
</tr>
<tr>
<td>Presence of others</td>
<td>13</td>
<td>82</td>
<td>1427</td>
</tr>
<tr>
<td>Anonymity</td>
<td>13</td>
<td>1493</td>
<td>16</td>
</tr>
<tr>
<td>Ability to backtrack</td>
<td>11</td>
<td>83</td>
<td>1428</td>
</tr>
</tbody>
</table>

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(university, hospital or other) in 331 cases (87%) and a private company in only 4%; therefore, this is not included in further analysis. The only additional consistently available information relating to cognitive burden was the number of items in a scale. Where this was not available from a paper it was gathered from elsewhere, giving information for 1456 (96%) comparisons. This is therefore included as a mediating factor in the meta-analysis. As the number of items per measure is highly skewed with a small number of outcomes having very large numbers of items, it was categorised by four percentile groups (Table 9).

An additional factor suggested in some reviews for technology-assisted data collection is timing of the study. When a technology is first introduced and is novel to the individuals within the study, greater differences may occur than once familiarisation has taken place. Date of data collection was poorly reported in studies, with 295 (77%) studies giving no indication of when their sample was recruited or data collected. Therefore, date of publication of the paper is used as an approximation to this. This distribution was highly skewed and, therefore, the data have been transformed.

**Assessment of mode effects on systematic bias**

Of the 1522 comparisons, 977 gave information to enable the calculation of a standardised mean difference. The mean within each mode was standardised and then the absolute mean difference between the two means taken as the summary statistic for this analysis. As this gives rise to an exponential distribution, the log of the absolute difference (plus 1) was taken for further analysis (Figure 9). This gives rise to a distribution that is left truncated at zero, but which, given the sample size, can be taken as normal for further analysis. This summary statistic captures the magnitude of differences between two modes on a standardised scale, so values can be interpreted as percentage differences.

Only 53% of studies contribute to this analysis; however, these represent 64% of the comparisons as those studies that report more comparisons are also reporting the data needed to calculate this summary statistic. As might be expected for this type of review, the level of clustering of outcome within studies overall is high [intracluster correlation (ICC) = 0.37], with studies considering within-person comparison of modes having a higher ICC (0.62) than between-group comparison studies (0.15). The ICC gives an indication of how similar the results are across the different outcomes measured within the same study.

A two-level linear regression model was fitted to the log of the absolute mean difference. The first model (Box 2) was fitted with the four main mode features representing the theoretical framework. Then the addition of other possible features was tested in model 2. The addition of date of publication as a mediating factor and interactions with the main mode features is included in model 3, as well as testing for the effect of study design. Model 4 is based on the

<table>
<thead>
<tr>
<th>No. of items</th>
<th>n</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>369</td>
<td>24.2</td>
</tr>
<tr>
<td>2–5</td>
<td>377</td>
<td>24.8</td>
</tr>
<tr>
<td>6–18</td>
<td>335</td>
<td>22.0</td>
</tr>
<tr>
<td>19+</td>
<td>375</td>
<td>24.6</td>
</tr>
<tr>
<td>Missing</td>
<td>66</td>
<td>4.3</td>
</tr>
<tr>
<td>Total</td>
<td>1522</td>
<td>100.0</td>
</tr>
</tbody>
</table>

**TABLE 9** Percentile groups for number of items within each measure
subset of comparisons with data on the number of items per measure. Each mode feature was coded to represent whether the two modes compared showed a difference on that feature or not, therefore a comparison of a face-to-face interview where the questions were read out loud and a telephone interview would have no difference in terms of sensory stimuli (both auditory), method of administration (both interviewer) or response (both verbal), but would show a difference in terms of use of a telephone and being online.

Fitting the model with absolute mean difference between the two mode features, we observed that, of the four main mode features, differences in administration (interviewer vs self) are highly significantly associated with larger differences between modes (Table 10). Differences in sensory stimuli are also significant, whereas the use of a computer or telephone has no impact on the magnitude of the difference between modes. On testing the additional possible features of mode (model 2), only the method of delivery approached significance and was, therefore, retained for further models. Model 3 shows that the date of publication is not associated with the magnitude of the difference and there are no significant interactions with the features associated with emerging technology (computer, telephone, sensory stimuli and delivery). The design of the study also had no impact on the model. Model 4 is fitted to the 941 comparisons in which data on the number of items within the measure are available. This shows a significant main effect with

---

**FIGURE 9** Histogram of logarithm of the absolute mean difference.

**BOX 2** Summary of models fitted

- Model 1: features from theoretical framework
- Model 2: model 1 + suggested other features
- Model 3: model 2 + date of publication and specified interactions
- Model 4: model 1 + anything significant from models 2 and 3 + cognitive burden (no. of items)
scales with more than one item associated with smaller differences between modes; however, there were no significant interactions with the mode features. This suggests that differences between modes reduce with increasing number of items and therefore cognitive burden.

### Assessment of mode effects on precision (variability)

Of the 1522 comparisons, 910 (60%) gave information on the SD or variance for each mode. One paper was excluded from this analysis because of the exceptionally large differences between variances (in excess of 100) suggestive of typographical errors. A two-level linear regression model was fitted as for the standardised mean difference (Table 11).

None of the mode features was associated with the size of the ratio of variances. The only variable that was significant was the design of the study, with between-group studies having greater differences between variances than within-group designs. This is as would be expected. No interactions were tested, as none of the main effects was significant.

---

**TABLE 10** Two-level regression models for absolute mean difference between two modes

<table>
<thead>
<tr>
<th>Variable</th>
<th>Model 1: (n=977)</th>
<th>Model 2: (n=977)</th>
<th>Model 3: (n=977)</th>
<th>Model 4: (n=941)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>B (SE) (p)-value</td>
<td>B (SE) (p)-value</td>
<td>B (SE) (p)-value</td>
<td>B (SE) (p)-value</td>
</tr>
<tr>
<td>Administration</td>
<td>(0.69\ (0.19)) (&lt;0.001)</td>
<td>(0.86\ (0.28)) (&lt;0.001)</td>
<td>(0.69\ (0.19)) (&lt;0.001)</td>
<td>(0.67\ (0.19)) (&lt;0.001)</td>
</tr>
<tr>
<td>Sensory stimuli</td>
<td>(-0.44\ (0.18)) (0.01)</td>
<td>(-0.37\ (0.19)) (0.05)</td>
<td>(-0.30\ (0.26)) (0.29)</td>
<td>(-0.43\ (0.18)) (0.02)</td>
</tr>
<tr>
<td>Computer</td>
<td>(-0.10\ (0.11)) (0.91)</td>
<td>(0.10\ (0.14)) (0.49)</td>
<td>(0.35\ (0.27)) (0.18)</td>
<td>(0.04\ (0.11)) (0.70)</td>
</tr>
<tr>
<td>Telephone</td>
<td>(0.09\ (0.08)) (0.29)</td>
<td>(-0.17\ (0.17)) (0.30)</td>
<td>(0.02\ (0.28)) (0.94)</td>
<td>(-0.09\ (0.10)) (0.39)</td>
</tr>
<tr>
<td>Delivery</td>
<td>(0.24\ (0.12)) (0.05)</td>
<td>(0.54\ (0.22)) (0.01)</td>
<td>(0.26\ (0.10)) (0.01)</td>
<td></td>
</tr>
<tr>
<td>Response</td>
<td>(-0.19\ (0.18)) (0.29)</td>
<td></td>
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</tr>
<tr>
<td>Online</td>
<td>(0.07\ (0.18)) (0.75)</td>
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<tr>
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<tr>
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<td>Date by sensory stimuli</td>
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<tr>
<td>Date by computer</td>
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</tr>
<tr>
<td>Date by telephone</td>
<td>(-0.03\ (0.13)) (0.44)</td>
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<td></td>
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</tr>
<tr>
<td>Date by delivery</td>
<td>(-0.18\ (0.11)) (0.09)</td>
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<tr>
<td>No. of items</td>
<td></td>
<td></td>
<td></td>
<td>(-0.21\ (0.10)) (0.01)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>Model 1: (n=977)</th>
<th>Model 2: (n=977)</th>
<th>Model 3: (n=977)</th>
<th>Model 4: (n=941)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Variances</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Level 2</td>
<td>(0.20\ (0.03)) (&lt;0.001)</td>
<td>(0.20\ (0.03)) (&lt;0.001)</td>
<td>(0.19\ (0.03)) (&lt;0.001)</td>
<td>(0.21\ (0.03)) (&lt;0.001)</td>
</tr>
<tr>
<td>Level 1</td>
<td>(0.38\ (0.02)) (&lt;0.001)</td>
<td>(0.37\ (0.02)) (&lt;0.001)</td>
<td>(0.37\ (0.02)) (&lt;0.001)</td>
<td>(0.37\ (0.02)) (&lt;0.001)</td>
</tr>
<tr>
<td>(-2LLH)</td>
<td>(2030.25) (\text{Ref.})</td>
<td>(2021.52)</td>
<td>(2016.48)</td>
<td>(0.03) (\text{n/a})</td>
</tr>
</tbody>
</table>

B, regression coefficient; LLH, log-likelihood; n/a, not applicable; Ref., reference.

\(\ast\) Not comparable to the other \(-2LLHs.\)

Bold text indicates \(p<0.05.\)
Assessment of mode effects on overall effect size

Data were available to calculate the ES for 912 comparisons (60%) (Table 12). The ES was calculated as the absolute difference between the means (raw) divided by the pooled SD.

Two-thirds of the ESs would be considered negligible (<0.2). This was highly skewed and, therefore, this was transformed prior to analysis (Figure 10).

A series of two-level linear regression models were then fitted as for the absolute mean difference (Table 13). The feature of administration is highly significant across all models, indicating a greater effect of this on the magnitude of differences between modes. Differences in sensory stimuli are of borderline significance in most models. Both the design of the study and the date of publication were significantly associated with ES. There were significant interactions between date of publication and computer and telephone usage. The numbers of items was significantly associated with ES, with smaller ESs for scales longer than one item. There was a significant interaction between this and the use of a computer.
Interpretability of results

The greatest impact of mode features is on the systematic bias in responses rather than the variability of responses. If we were to take a hypothetical example for a measure, such as a subscale with two to five items from the SF-36 scored from 0 to 100, then the impact of the two significant variables ‘administration’ and ‘sensory stimuli’ on the absolute mean difference (systematic bias) is shown in Table 14, in terms of the predicted absolute mean differences.

This is what we would predict in terms of absolute mean difference if we were to design a factorial trial with two measurements carried out on each participant. However, if we want to relate this to mean difference (instead of absolute mean difference), we need to make some assumptions. It is reasonable to assume that, in the absence of any differences in mode or features causing biased responding, that the upper right-hand cell represents a half-normal distribution centred on zero. This relates to a normal distribution for differences with a mean of zero and an estimated SD of approximately ‘5’. The most commonly occurring combination of these two mode features is to have both a difference in administration and a difference in sensory stimuli, which, for a measure such as the SF-36, would result in an expected bias of 0.85 units, assuming no impact on the SD.

### TABLE 12 Effect sizes in categories

<table>
<thead>
<tr>
<th>ES</th>
<th>No. (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.0–0.1999</td>
<td>604 (66.2)</td>
</tr>
<tr>
<td>0.2–0.3999</td>
<td>176 (11.3)</td>
</tr>
<tr>
<td>0.4–0.5999</td>
<td>67 (7.3)</td>
</tr>
<tr>
<td>0.6–0.9999</td>
<td>50 (5.5)</td>
</tr>
<tr>
<td>1.0–1.9999</td>
<td>12 (1.3)</td>
</tr>
<tr>
<td>≥ 2.0 and greater</td>
<td>3 (0.3)</td>
</tr>
</tbody>
</table>

### FIGURE 10 Distribution of transformed ES.
<table>
<thead>
<tr>
<th>Variable</th>
<th>Model 1: $n = 912$</th>
<th></th>
<th>Model 2: $n = 912$</th>
<th></th>
<th>Model 3: $n = 912$</th>
<th></th>
<th>Model 4: $n = 888$</th>
<th></th>
<th>Model 5: $n = 888$</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>B (SE) p-value</td>
<td></td>
<td>B (SE) p-value</td>
<td></td>
<td>B (SE) p-value</td>
<td></td>
<td>B (SE) p-value</td>
<td></td>
<td>B (SE) p-value</td>
<td></td>
</tr>
<tr>
<td>Administration</td>
<td>0.57 (0.20) 0.003</td>
<td></td>
<td>0.71 (0.30) 0.02</td>
<td></td>
<td>0.57 (0.19) 0.003</td>
<td></td>
<td>0.56 (0.19) 0.003</td>
<td></td>
<td>0.57 (0.19) 0.003</td>
<td></td>
</tr>
<tr>
<td>Sensory stimuli</td>
<td>–0.38 (0.19) 0.05</td>
<td>0.08</td>
<td>–0.27 (0.23) 0.23</td>
<td></td>
<td>–0.39 (0.18) 0.03</td>
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<td>–0.39 (0.18) 0.03</td>
<td></td>
<td>–0.03 (0.25) 0.89</td>
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</tr>
<tr>
<td>Computer</td>
<td>0.14 (0.10) 0.16</td>
<td></td>
<td>0.24 (0.13) 0.06</td>
<td></td>
<td>0.57 (0.18) 0.001</td>
<td></td>
<td>0.50 (0.15) 0.001</td>
<td></td>
<td>–0.03 (0.25) 0.89</td>
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</tr>
<tr>
<td>Telephone</td>
<td>0.12 (0.08) 0.14</td>
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<td>0.13 (0.16) 0.40</td>
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<td>0.83 (0.27) 0.002</td>
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<td>0.74 (0.25) 0.003</td>
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<td>0.67 (0.25) 0.008</td>
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<td>Delivery</td>
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<td>0.10 (0.19) 0.58</td>
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<td>0.16 (0.10) 0.11</td>
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<td>0.15 (0.10) 0.13</td>
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<td>Response</td>
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<td>Online</td>
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<td>Presence of interviewer</td>
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<td>Date of publication</td>
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<tr>
<td>Date by sensory stimuli</td>
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<tr>
<td>Date by computer</td>
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<td>0.29 &lt;0.001</td>
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<td>0.26 &lt;0.001</td>
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<td>0.25 &lt;0.001</td>
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<td>0.25 &lt;0.001</td>
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<tr>
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<td>0.26 &lt;0.001</td>
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<td>0.26 &lt;0.001</td>
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<td>0.25 &lt;0.001</td>
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<td>0.25 &lt;0.001</td>
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<tr>
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<td>1648.25 0.08</td>
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<td>1565.82 Ref.</td>
<td></td>
<td>1556.36 0.02</td>
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</tr>
</tbody>
</table>

B, regression coefficient; LLH, log-likelihood; n/a, not applicable; Ref., reference.

a Not comparable to the other –2LLHs.
The most frequently occurring individual outcome measure within the studies included was the SF-36 health survey. The SF-36 consists of eight aggregate scale scores. Each scale is directly transformed into a 0–100 scale on the assumption that each contributing item carries equal weight. The eight scales are vitality, physical functioning, bodily pain, general health, role physical, role emotional, role mental and mental health.

Seventeen studies published between 1994 and 2003 used SF-36. Not all studies reported all subscales. The impact of the different modes of using SF-36 was assessed using weighted pooled measures of agreement for within-subject comparisons and random-effects meta-analysis for between-subject comparison. There were seven studies that provided between-subject comparisons only, eight studies that provided within-subject comparisons only and two studies that contributed data to both analyses. Table 15 summarises the information available from each study.

### Between-subject comparisons

Eight studies had some data available that could contribute to the meta-analysis. One of these (Amodei et al.) was a comparison of an interview in which the interviewer asked the questions and recorded the response to one in which the interviewer asked the questions and the responder confidentially recorded their own response. This mode comparison does not reflect a difference on one of the four mode features and, therefore, has not been included in the subsequent analysis. One of the crossover studies (Lyons et al.) provided only mean scores at the first time point and, therefore, could not be included in this analysis.

### Within-subject comparisons

There was a greater variety in the statistical approaches taken to analysis in the within-subject studies, and the data presented that could contribute to the pooled analysis were limited. Studies that did not give information on mean differences and SDs tended to report correlations. The available studies have been combined to give pooled estimators of mean difference with 95% CIs and pooled limits of agreement.

### Mode feature: computer

Only two of the between-subjects studies contributed to the analysis of the computerisation mode feature. The results of the meta-analysis for each subscale of the SF-36 can be seen in Figures 11–18 (forest plots in order of magnitude of pooled difference).

Role emotional, social functioning and mental health (see Figures 11–13) all suggest that significantly higher scores are achieved with computers than without, with mean differences of between four and eight points on the scale. It should be noted that as the Perkins and Sanson-Fisher study is 10 times the size of the Saleh et al. study, it dominates the pooled estimator.

<table>
<thead>
<tr>
<th>Differences in administration</th>
<th>No</th>
<th>Yes</th>
</tr>
</thead>
<tbody>
<tr>
<td>Difference in sensory stimuli</td>
<td>2.07</td>
<td>5.01</td>
</tr>
<tr>
<td></td>
<td>1.01</td>
<td>2.92</td>
</tr>
</tbody>
</table>

**TABLE 14** Predicted absolute mean differences from Model 4

**Meta-analysis of Short Form questionnaire-36 items**

The most frequently occurring individual outcome measure within the studies included was the SF-36 health survey. The SF-36 consists of eight aggregate scale scores. Each scale is directly transformed into a 0–100 scale on the assumption that each contributing item carries equal weight. The eight scales are vitality, physical functioning, bodily pain, general health, role physical, role emotional, role mental and mental health.

Seventeen studies published between 1994 and 2003 used SF-36. Not all studies reported all subscales. The impact of the different modes of using SF-36 was assessed using weighted pooled measures of agreement for within-subject comparisons and random-effects meta-analysis for between-subject comparison. There were seven studies that provided between-subject comparisons only, eight studies that provided within-subject comparisons only and two studies that contributed data to both analyses. Table 15 summarises the information available from each study.

### Between-subject comparisons

Eight studies had some data available that could contribute to the meta-analysis. One of these (Amodei et al.) was a comparison of an interview in which the interviewer asked the questions and recorded the response to one in which the interviewer asked the questions and the responder confidentially recorded their own response. This mode comparison does not reflect a difference on one of the four mode features and, therefore, has not been included in the subsequent analysis. One of the crossover studies (Lyons et al.) provided only mean scores at the first time point and, therefore, could not be included in this analysis.

### Within-subject comparisons

There was a greater variety in the statistical approaches taken to analysis in the within-subject studies, and the data presented that could contribute to the pooled analysis were limited. Studies that did not give information on mean differences and SDs tended to report correlations. The available studies have been combined to give pooled estimators of mean difference with 95% CIs and pooled limits of agreement.

### Mode feature: computer

Only two of the between-subjects studies contributed to the analysis of the computerisation mode feature. The results of the meta-analysis for each subscale of the SF-36 can be seen in Figures 11–18 (forest plots in order of magnitude of pooled difference).

Role emotional, social functioning and mental health (see Figures 11–13) all suggest that significantly higher scores are achieved with computers than without, with mean differences of between four and eight points on the scale. It should be noted that as the Perkins and Sanson-Fisher study is 10 times the size of the Saleh et al. study, it dominates the pooled estimator.
Only one of the within-subjects studies\textsuperscript{124} provided data on this mode feature. The results for the Ryan \textit{et al.} study are given in Table 16.

The only outcome for which there was a significant difference was for 'social functioning', with higher scores for those not using a computer. This is contrary to the findings in the
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
</tr>
<tr>
<td>Perkins 1998(^{14})</td>
<td>88.76</td>
<td>27.99</td>
<td>421</td>
</tr>
<tr>
<td>Saleh 2002(^{15})</td>
<td>56.9</td>
<td>44.2</td>
<td>41</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td></td>
<td></td>
<td>462</td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 0.53$, df = 1 ($p = 0.47$); $F = 0\%$
Test for overall effect: $z = 3.88$ ($p = 0.0001$)

FIGURE 11 Meta-analysis SF-36 – role emotional\*, *, $p < 0.05$.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
</tr>
<tr>
<td>Perkins 1998(^{14})</td>
<td>89.25</td>
<td>22.6</td>
<td>421</td>
</tr>
<tr>
<td>Saleh 2002(^{15})</td>
<td>65.4</td>
<td>27.4</td>
<td>39</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td></td>
<td></td>
<td>460</td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 0.52$, df = 1 ($p = 0.47$); $F = 0\%$
Test for overall effect: $z = 3.00$ ($p = 0.003$)

FIGURE 12 Meta-analysis SF-36 – social functioning\*, *, $p < 0.05$. 
### Study or subgroup

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Perkins 1998¹¹⁴</td>
<td>4.05 (1.77 to 6.33)</td>
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</tr>
<tr>
<td>Saleh 2002¹¹⁵</td>
<td>2.20 (–6.42 to 10.82)</td>
<td></td>
</tr>
<tr>
<td><strong>Total</strong> (95% CI)</td>
<td><strong>3.93 (1.72 to 6.13)</strong></td>
<td><strong>3.03 (–1.66 to 7.73)</strong></td>
</tr>
</tbody>
</table>

#### FIGURE 13 Meta-analysis SF-36 – mental health. *, *p* < 0.05.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
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<tbody>
<tr>
<td>Perkins 1998¹¹⁴</td>
<td>80.34</td>
<td>35.7</td>
<td>421</td>
<td>76.77</td>
<td>36.44</td>
<td>418</td>
<td>92.4</td>
<td>3.57 (–1.31 to 8.45)</td>
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<tr>
<td>Saleh 2002¹¹⁵</td>
<td>27.4</td>
<td>40.2</td>
<td>41</td>
<td>30.9</td>
<td>39.9</td>
<td>44</td>
<td>7.6</td>
<td>–3.50 (–20.54 to 13.54)</td>
<td></td>
</tr>
<tr>
<td><strong>Total</strong> (95% CI)</td>
<td><strong>462</strong></td>
<td>100.0</td>
<td><strong>462</strong></td>
<td>100.0</td>
<td><strong>3.03</strong> (–1.66 to 7.73)</td>
<td><strong>3.03</strong> (–1.66 to 7.73)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

#### FIGURE 14 Meta-analysis SF-36 – role physical. *, *p* < 0.05.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Perkins 1998</td>
<td>63.22</td>
<td>60.57</td>
<td>2.65 (−0.35 to 5.65)</td>
</tr>
<tr>
<td>Saleh 2002</td>
<td>48.7</td>
<td>49</td>
<td>−0.30 (−8.71 to 8.11)</td>
</tr>
</tbody>
</table>

Total (95% CI) 460 463 100.0 2.32 (−0.50 to 5.14)

Heterogeneity: $\tau^2 = 0.00; \chi^2 = 0.42, df = 1 (p = 0.52); F = 0%$
Test for overall effect: $z = 1.61 (p = 0.11)$

FIGURE 15 Meta-analysis SF-36 – vitality.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Perkins 1998</td>
<td>72.84</td>
<td>71.01</td>
<td>1.83 (−1.17 to 4.83)</td>
</tr>
<tr>
<td>Saleh 2002</td>
<td>56</td>
<td>56.7</td>
<td>−0.70 (−9.60 to 8.20)</td>
</tr>
</tbody>
</table>

Total (95% CI) 457 463 100.0 1.57 (−1.27 to 4.42)

Heterogeneity: $\tau^2 = 0.00; \chi^2 = 0.28, df = 1 (p = 0.60); F = 0%$
Test for overall effect: $z = 1.08 (p = 0.28)$

FIGURE 16 Meta-analysis SF-36 – general health perception.
### FIGURE 17 Meta-analysis SF-36 – physical functioning.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Perkins 1998(^\text{114})</td>
<td>82.64 (24.16) 421</td>
<td>81.35 (22.8) 418</td>
<td>90.5</td>
<td>1.29 (−1.89 to 4.47)</td>
</tr>
<tr>
<td>Saleh 2002(^\text{115})</td>
<td>40.5 (19.5) 41</td>
<td>38.9 (26.3) 44</td>
<td>9.5</td>
<td>1.60 (−8.20 to 11.40)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>462</td>
<td>462</td>
<td>100.0</td>
<td>1.32 (−1.70 to 4.34)</td>
</tr>
</tbody>
</table>

Heterogeneity: \( \tau^2 = 0.00; \chi^2 = 0.00, df = 1 (p = 0.95); F = 0\%

Test for overall effect: \( z = 0.86 (p = 0.39) \)

---

### FIGURE 18 Meta-analysis SF-36 – bodily pain.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computerised</th>
<th>Not computerised</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Perkins 1998(^\text{114})</td>
<td>77.75 (27.64) 421</td>
<td>73.73 (24.41) 421</td>
<td>59.2</td>
<td>4.02 (0.50 to 7.54)</td>
</tr>
<tr>
<td>Saleh 2002(^\text{115})</td>
<td>36.2 (17.6) 39</td>
<td>40.1 (17.4) 44</td>
<td>40.8</td>
<td>−3.90 (−11.45 to 3.65)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>460</td>
<td>465</td>
<td>100.0</td>
<td>0.79 (−6.84 to 8.42)</td>
</tr>
</tbody>
</table>

Heterogeneity: \( \tau^2 = 22.34; \chi^2 = 3.47, df = 1 (p = 0.06); I^2 = 71\%

Test for overall effect: \( z = 0.20 (p = 0.84) \)
between-group analysis, but other than social and physical functioning all results from this study go in the same direction as those from the between-group meta-analysis. It should be noted that the limits of agreement are very wide for all outcomes – this indicates that there could be considerable differences at an individual level. Although this may be of less concern to researchers, who are usually comparing groups, this would be much more of an issue if different modes were being used in clinical care and decisions on an individual basis.

**Mode feature: administration and sensory stimuli**

Seven between-subject studies compared modes in which there was a difference in administration. All of these also had a difference in sensory stimuli, with auditory stimuli with interviewer administration and visual stimuli with self. Of these, five studies provided data that could contribute to a meta-analysis. The results of the meta-analysis for each subscale of the SF-36 can be seen in Figures 19–26 (forest plots in order of magnitude of pooled difference).

None of the scales show a significant difference between interviewer and self-administration, although all are in the direction of self-completion giving rise to higher scores. However, there was a high degree of heterogeneity between studies. The Jones et al. study used the Veteran’s SF-36, which was developed from the SF-36 to be specifically used in the Veteran’s Health Administration. Particular changes were made to the two subscales measuring role (physical and emotional) during the development process (see Figures 22 and 25). If the Jones et al. study were to be excluded from the meta-analysis, the greatest impact would be on the effect for the ‘role emotional’ subscale, which would become significantly higher with interviewer administration [6.82 (95% CI 2.61 to 11.03)]; however, high levels of heterogeneity still remain. For the ‘role physical’ subscale, the effect changed sign, but was still not significant [2.24 (95% CI –3.28 to 7.76)]. For the other scales, three of the remaining six would also become positive, indicating higher scores for interviewer administration.

Two studies provided data on differences in administration from the within-subject studies. The pooled estimators of effect can be seen in Table 17.

The pooled data from these two studies suggest higher scores for interviewer administration for all subscales, with all but bodily pain and vitality being significant. The impact on the two role subscales is in the order of 10 points; however, this is based on a total of only 250 patients.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Bowling 1999104</td>
<td>89.6</td>
<td>19.3</td>
<td>2025</td>
<td>88.4</td>
</tr>
<tr>
<td>Jones 2001112</td>
<td>36.86</td>
<td>27.84</td>
<td>1591</td>
<td>50.78</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>82.64</td>
<td>24.16</td>
<td>421</td>
<td>81.35</td>
</tr>
<tr>
<td>Unruh 2003118</td>
<td>41.5</td>
<td>26.2</td>
<td>426</td>
<td>51.5</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>53.6</td>
<td>27.2</td>
<td>136</td>
<td>51.3</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>4599</td>
<td>11,456</td>
<td>100.0</td>
<td><strong>-4.17 (-11.98 to 3.64)</strong></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 73.43; \chi^2 = 205.43, \text{df } = 4 (p < 0.00001); I^2 = 98\%$

Test for overall effect: $z = 1.05 (p = 0.30)$

---

**FIGURE 19** Meta-analysis SF-36 – physical functioning.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Bowling 1999104</td>
<td>74</td>
<td>21.9</td>
<td>2017</td>
<td>73.5</td>
</tr>
<tr>
<td>Jones 2001112</td>
<td>33.18</td>
<td>21.84</td>
<td>1591</td>
<td>41.17</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>72.84</td>
<td>22.7</td>
<td>421</td>
<td>71.01</td>
</tr>
<tr>
<td>Unruh 2003118</td>
<td>44.6</td>
<td>21.1</td>
<td>422</td>
<td>47.4</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>35.4</td>
<td>22</td>
<td>136</td>
<td>40.3</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>4587</td>
<td>11,639</td>
<td>100.0</td>
<td><strong>-2.52 (-6.90 to 1.85)</strong></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 21.70; \chi^2 = 78.94, \text{df } = 4 (p < 0.00001); I^2 = 95\%$

Test for overall effect: $z = 1.13 (p = 0.26)$

**FIGURE 20** Meta-analysis SF-36 – general health perception.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean (SD)</td>
<td>Mean (SD)</td>
<td>Weight (%) IV, Random, 95% CI</td>
</tr>
<tr>
<td>Bowling 1999</td>
<td>89 (20.8)</td>
<td>88 (19.5)</td>
<td>9124 (22.1) 1.00 (0.01 to 1.99)</td>
</tr>
<tr>
<td>Jones 2001</td>
<td>43.37 (30.92)</td>
<td>55.08 (34.41)</td>
<td>1659 (21.6) -11.71 (-13.96 to -9.46)</td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>89.25 (22.6)</td>
<td>84.59 (21.2)</td>
<td>418 (21.2) 4.66 (1.70 to 7.62)</td>
</tr>
<tr>
<td>Unruh 2003</td>
<td>70.8 (29.5)</td>
<td>70.9 (25.5)</td>
<td>550 (20.8) -0.10 (-3.62 to -3.42)</td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>62.2 (29.4)</td>
<td>69.8 (26.6)</td>
<td>36 (14.2) -7.60 (-17.60 to 2.40)</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>4593 (11,787)</td>
<td>100.0 (11,787)</td>
<td>-2.42 (-8.68 to 3.83)</td>
</tr>
</tbody>
</table>

Heterogeneity: $I^2 = 45.68; \chi^2 = 117.95, df = 4 (p < 0.00001); f^2 = 97\%$
Test for overall effect: $z = 0.76 (p = 0.45)\$

**FIGURE 21** Meta-analysis SF-36 – social functioning.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean (SD)</td>
<td>Mean (SD)</td>
<td>Weight (%) IV, Random, 95% CI</td>
</tr>
<tr>
<td>Bowling 1999</td>
<td>84.2 (32.7)</td>
<td>85.8 (29.9)</td>
<td>9058 (22.1) -1.60 (-3.15 to -0.05)</td>
</tr>
<tr>
<td>Jones 2001</td>
<td>18.92 (30.96)</td>
<td>35.37 (40.98)</td>
<td>1659 (21.9) -16.45 (-18.94 to -13.96)</td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>80.34 (35.7)</td>
<td>76.77 (36.44)</td>
<td>418 (20.9) 3.57 (-1.31 to 8.45)</td>
</tr>
<tr>
<td>Unruh 2003</td>
<td>50.4 (40.5)</td>
<td>42.2 (40.5)</td>
<td>538 (20.7) 8.20 (3.05 to 13.35)</td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>30.7 (36)</td>
<td>34.6 (38.9)</td>
<td>36 (14.5) -3.90 (-17.97 to 10.17)</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>4592 (11,709)</td>
<td>100.0 (11,709)</td>
<td>-2.07 (-11.13 to 7.00)</td>
</tr>
</tbody>
</table>

Heterogeneity: $I^2 = 96.26; \chi^2 = 135.65, df = 4 (p < 0.00001); f^2 = 97\%$
Test for overall effect: $z = 0.45 (p = 0.65)\$

**FIGURE 22** Meta-analysis SF-36 – role physical.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Weight (%)</th>
<th>IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bowling 1999104</td>
<td>64.7</td>
<td>20.8</td>
<td>2018</td>
<td>61.1</td>
<td>19.6</td>
</tr>
<tr>
<td>Jones 2001112</td>
<td>31.2</td>
<td>22.6</td>
<td>1591</td>
<td>35.4</td>
<td>26.12</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>63.2</td>
<td>22.7</td>
<td>421</td>
<td>60.57</td>
<td>21.48</td>
</tr>
<tr>
<td>Unruh 2003115</td>
<td>47.2</td>
<td>24.7</td>
<td>425</td>
<td>51.6</td>
<td>19.3</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>35.3</td>
<td>24</td>
<td>136</td>
<td>43.2</td>
<td>24</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>4591</strong></td>
<td></td>
<td><strong>11,656</strong></td>
<td><strong>100.0</strong></td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 22.84; \chi^2 = 81.57, df = 4 (p < 0.00001); F = 95%$

Test for overall effect: $z = 0.64 (p = 0.53)$

**FIGURE 23** Meta-analysis SF-36 – vitality.

---

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Interviewer</th>
<th>Self</th>
<th>Weight (%)</th>
<th>IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bowling 1999104</td>
<td>76.6</td>
<td>18.3</td>
<td>2019</td>
<td>73.8</td>
<td>17.2</td>
</tr>
<tr>
<td>Jones 2001112</td>
<td>53.84</td>
<td>26.46</td>
<td>1591</td>
<td>58.76</td>
<td>28.13</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>80.38</td>
<td>17.42</td>
<td>421</td>
<td>76.33</td>
<td>16.28</td>
</tr>
<tr>
<td>Unruh 2003115</td>
<td>71.1</td>
<td>21.5</td>
<td>424</td>
<td>71.7</td>
<td>17.8</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>62.5</td>
<td>25.2</td>
<td>136</td>
<td>75</td>
<td>16.5</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>4591</strong></td>
<td></td>
<td><strong>11,589</strong></td>
<td><strong>100.0</strong></td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 18.64; \chi^2 = 76.77, df = 4 (p < 0.00001); F = 95%$

Test for overall effect: $z = 0.69 (p = 0.49)$

**FIGURE 24** Meta-analysis SF-36 – mental health.
### Meta-analysis SF-36 – role emotional.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Weight (%)</th>
<th>IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bowling 1999</td>
<td>82.5</td>
<td>24.8</td>
<td>2022</td>
<td>81.5</td>
<td>21.6</td>
<td>10105</td>
<td>23.9</td>
<td>1.00 (-0.16 to 2.16)</td>
<td></td>
</tr>
<tr>
<td>Jones 2001</td>
<td>39.15</td>
<td>27.44</td>
<td>1591</td>
<td>45.77</td>
<td>29.88</td>
<td>1659</td>
<td>23.2</td>
<td>-6.62 (-8.59 to -4.65)</td>
<td></td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>77.75</td>
<td>27.64</td>
<td>421</td>
<td>73.73</td>
<td>24.41</td>
<td>421</td>
<td>21.0</td>
<td>4.02 (0.50 to 7.54)</td>
<td></td>
</tr>
<tr>
<td>Unruh 2003</td>
<td>64.8</td>
<td>29.4</td>
<td>426</td>
<td>62.2</td>
<td>26.3</td>
<td>548</td>
<td>20.9</td>
<td>2.60 (-0.96 to 6.16)</td>
<td></td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>43.3</td>
<td>27.3</td>
<td>136</td>
<td>46.7</td>
<td>26.4</td>
<td>36</td>
<td>10.9</td>
<td>-3.40 (-13.17 to 6.37)</td>
<td></td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>4596</td>
<td></td>
<td>12,769</td>
<td>100.0</td>
<td></td>
<td>-0.28</td>
<td>-4.63 to 4.07</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Heterogeneity:** $\chi^2 = 20.20; \chi^2 = 52.99, df = 4 (p < 0.0001); I^2 = 92%$

Test for overall effect: $z = 0.13 (p = 0.90)$

### FIGURE 25
Meta-analysis SF-36 – role emotional.

### Meta-analysis SF-36 – bodily pain.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Mean</th>
<th>SD</th>
<th>Total</th>
<th>Weight (%)</th>
<th>IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bowling 1999</td>
<td>88</td>
<td>29.1</td>
<td>1919</td>
<td>82.9</td>
<td>31.8</td>
<td>8067</td>
<td>21.1</td>
<td>5.10 (3.62 to 6.58)</td>
<td></td>
</tr>
<tr>
<td>Jones 2001</td>
<td>32.03</td>
<td>38.47</td>
<td>1591</td>
<td>54.48</td>
<td>43.91</td>
<td>1659</td>
<td>21.0</td>
<td>-22.45 (-25.29 to -19.61)</td>
<td></td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>88.76</td>
<td>27.99</td>
<td>421</td>
<td>80.36</td>
<td>33.47</td>
<td>418</td>
<td>20.8</td>
<td>8.40 (4.22 to 12.58)</td>
<td></td>
</tr>
<tr>
<td>Unruh 2003</td>
<td>70.7</td>
<td>40.2</td>
<td>426</td>
<td>59.2</td>
<td>43.4</td>
<td>534</td>
<td>20.5</td>
<td>11.50 (6.20 to 16.80)</td>
<td></td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>54.7</td>
<td>43.9</td>
<td>136</td>
<td>63.9</td>
<td>43.2</td>
<td>310</td>
<td>16.6</td>
<td>-9.20 (-25.12 to 6.72)</td>
<td></td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>4493</td>
<td></td>
<td>10,714</td>
<td>100.0</td>
<td></td>
<td>-1.06</td>
<td>-15.07 to 12.96</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Heterogeneity:** $\chi^2 = 320.74, df = 4 (p < 0.00001); I^2 = 99%$

Test for overall effect: $z = 0.15 (p = 0.88)$

### FIGURE 26
Meta-analysis SF-36 – bodily pain.
Mode feature: telephone

Three between-subject studies\textsuperscript{112,114,118} provided data for consideration of the impact of the telephone mode feature. The results of the meta-analysis for each subscale of the SF-36 can be seen in Figures 27–34 (forest plots by order of magnitude of pooled difference).

All of the subscales had differences in the direction of giving higher scores without a telephone. As for the previous analysis, there were high levels of heterogeneity. The two largest effects were for the 'role physical' and 'role emotional' scales (see Figures 30 and 31). Excluding the Jones \textit{et al.} study\textsuperscript{112} from these (as it was using Veteran's SF-36) would considerably reduce the estimated mean difference (to –0.77 and 1.45, respectively).

Two studies\textsuperscript{118,123} provided data on differences in telephone administration from the within-subject studies. The pooled estimators of effect can be seen in Table 18.

All of the subscales except 'vitality' show a mean difference in the direction of higher scores without a telephone, which is consistent with the results from the between-group studies. Only 'role physical' shows a significant difference.

**Meta-analysis of the Minnesota Multiphasic Personality Inventory**

The second most frequently occurring measure from the studies included was the MMPI.\textsuperscript{128} The MMPI was developed in the 1930s at Minnesota University as a comprehensive personality test that could be used to detect psychiatric problems. The MMPI consists of 14 scaled scores. Ten clinical scales are included to indicate different psychiatric conditions (hypochondriasis,

<table>
<thead>
<tr>
<th>Outcome</th>
<th>n</th>
<th>Mean diff.</th>
<th>SD (diff.)</th>
<th>95% CI</th>
<th>95% limits of agreement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Role emotional</td>
<td>250</td>
<td>12.8</td>
<td>41.6</td>
<td>7.6 to 17.9</td>
<td>–68.7 to 94.2</td>
</tr>
<tr>
<td>Role physical</td>
<td>250</td>
<td>10.0</td>
<td>31.8</td>
<td>5.8 to 14.2</td>
<td>–56.5 to 76.5</td>
</tr>
<tr>
<td>Social functioning</td>
<td>250</td>
<td>5.5</td>
<td>22.3</td>
<td>2.7 to 8.2</td>
<td>–38.3 to 49.3</td>
</tr>
<tr>
<td>Physical functioning</td>
<td>250</td>
<td>5.2</td>
<td>18.1</td>
<td>3.0 to 7.5</td>
<td>–30.3 to 40.7</td>
</tr>
<tr>
<td>General health</td>
<td>250</td>
<td>3.5</td>
<td>14.0</td>
<td>1.7 to 5.2</td>
<td>–24.0 to 30.9</td>
</tr>
<tr>
<td>Mental health</td>
<td>250</td>
<td>2.9</td>
<td>15.0</td>
<td>1.0 to 4.7</td>
<td>–26.5 to 32.2</td>
</tr>
<tr>
<td>Bodily pain</td>
<td>250</td>
<td>1.5</td>
<td>21.2</td>
<td>–1.1 to 4.2</td>
<td>–40.0 to 43.1</td>
</tr>
<tr>
<td>Vitality</td>
<td>250</td>
<td>0.7</td>
<td>16.9</td>
<td>–1.4 to 2.8</td>
<td>–32.3 to 33.8</td>
</tr>
</tbody>
</table>

diff., difference.

<table>
<thead>
<tr>
<th>Outcome</th>
<th>n</th>
<th>Mean diff.</th>
<th>SD (diff.)</th>
<th>95% CI</th>
<th>95% limits of agreement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Role physical</td>
<td>73</td>
<td>–6.1</td>
<td>22.6</td>
<td>–11.3 to –0.9</td>
<td>–50.4 to 38.2</td>
</tr>
<tr>
<td>Social functioning</td>
<td>73</td>
<td>–4.1</td>
<td>25.9</td>
<td>–10.0 to 1.9</td>
<td>–54.8 to 46.7</td>
</tr>
<tr>
<td>Role emotional</td>
<td>73</td>
<td>–3.9</td>
<td>25.9</td>
<td>–9.8 to 2.0</td>
<td>–54.6 to 46.8</td>
</tr>
<tr>
<td>Physical functioning</td>
<td>73</td>
<td>–2.1</td>
<td>12.6</td>
<td>–5.0 to 0.8</td>
<td>–26.7 to 22.5</td>
</tr>
<tr>
<td>Bodily pain</td>
<td>73</td>
<td>–0.7</td>
<td>15.1</td>
<td>–4.1 to 2.8</td>
<td>–30.3 to 29.0</td>
</tr>
<tr>
<td>General health</td>
<td>73</td>
<td>–0.3</td>
<td>13.6</td>
<td>–3.4 to 2.8</td>
<td>–27.0 to 26.4</td>
</tr>
<tr>
<td>Mental health</td>
<td>73</td>
<td>–0.2</td>
<td>10.6</td>
<td>–2.6 to 2.2</td>
<td>–20.9 to 20.6</td>
</tr>
<tr>
<td>Vitality</td>
<td>73</td>
<td>0.8</td>
<td>14.9</td>
<td>–2.7 to 4.2</td>
<td>–28.4 to 29.9</td>
</tr>
</tbody>
</table>
### FIGURE 27 Meta-analysis SF-36 – role physical.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone</th>
<th>No telephone</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Heterogeneity: $\tau^2 = 162.80$; $\chi^2 = 51.35$, df = 2 ($p &lt; 0.00001$); $I^2 = 96%$</th>
<th>Test for overall effect: $z = 1.01$ ($p = 0.31$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001112</td>
<td>18.92</td>
<td>30.96</td>
<td>1591</td>
<td>35.6</td>
<td>16.45 ($-18.94$ to $-13.96$)</td>
</tr>
<tr>
<td>Perkins 199814</td>
<td>80.34</td>
<td>35.7</td>
<td>421</td>
<td>34.6</td>
<td>3.57 ($-1.31$ to $8.45$)</td>
</tr>
<tr>
<td>Weinberger 199618</td>
<td>23.9</td>
<td>32.5</td>
<td>47</td>
<td>29.7</td>
<td>$-10.50$ ($-21.93$ to $0.93$)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td>100.0</td>
<td><strong>-7.74</strong> ($-22.74$ to $7.25$)</td>
<td></td>
</tr>
</tbody>
</table>

### FIGURE 28 Meta-analysis SF-36 – role emotional.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone</th>
<th>No telephone</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Heterogeneity: $\tau^2 = 430.78$; $\chi^2 = 143.79$, df = 2 ($p &lt; 0.00001$); $I^2 = 99%$</th>
<th>Test for overall effect: $z = 0.61$ ($p = 0.54$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001112</td>
<td>32.03</td>
<td>38.47</td>
<td>1591</td>
<td>34.7</td>
<td>$-22.45$ ($-25.29$ to $-19.61$)</td>
</tr>
<tr>
<td>Perkins 199814</td>
<td>88.76</td>
<td>27.99</td>
<td>421</td>
<td>34.5</td>
<td>8.40 ($4.22$ to $12.58$)</td>
</tr>
<tr>
<td>Weinberger 199618</td>
<td>50.4</td>
<td>43.3</td>
<td>47</td>
<td>30.9</td>
<td>$-8.50$ ($-23.08$ to $6.08$)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td>100.0</td>
<td><strong>-7.51</strong> ($-31.52$ to $16.50$)</td>
<td></td>
</tr>
</tbody>
</table>
### Table: SF-36 – Social Functioning

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone Mean (SD)</th>
<th>No telephone Mean (SD)</th>
<th>Weight (%)</th>
<th>Mean difference (IV, Random, 95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001</td>
<td>43.37 (30.92)</td>
<td>55.08 (34.41)</td>
<td>35.3</td>
<td>-11.71 (-13.96 to -9.46)</td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>80.25 (22.6)</td>
<td>84.59 (21.2)</td>
<td>35.0</td>
<td>4.66 (1.70 to 7.62)</td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>54.5 (29.2)</td>
<td>67.3 (28.7)</td>
<td>29.6</td>
<td>-12.80 (-22.55 to -3.05)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td><strong>100.0</strong></td>
<td><strong>-6.30 (-19.17 to 6.58)</strong></td>
</tr>
</tbody>
</table>

**FIGURE 29** Meta-analysis SF-36 – social functioning.

### Table: SF-36 – Physical Functioning

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone Mean (SD)</th>
<th>No telephone Mean (SD)</th>
<th>Weight (%)</th>
<th>Mean difference (IV, Random, 95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001</td>
<td>36.86 (27.84)</td>
<td>50.78 (31.08)</td>
<td>35.6</td>
<td>-13.92 (-13.95 to -11.89)</td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>82.64 (24.16)</td>
<td>81.35 (22.8)</td>
<td>35.1</td>
<td>1.29 (-1.89 to 4.47)</td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>49 (27.9)</td>
<td>54.7 (27.7)</td>
<td>29.3</td>
<td>-5.70 (-15.04 to 3.64)</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td><strong>100.0</strong></td>
<td><strong>-6.18 (-17.96 to 5.61)</strong></td>
</tr>
</tbody>
</table>

**FIGURE 30** Meta-analysis SF-36 – physical functioning.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone</th>
<th>No telephone</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001112</td>
<td>33.18</td>
<td>21.84</td>
<td>41.17</td>
<td>26.31</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>72.84</td>
<td>22.7</td>
<td>71.01</td>
<td>21.64</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>31.9</td>
<td>22.4</td>
<td>38.1</td>
<td>21.6</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td><strong>100.0</strong></td>
<td><strong>-4.01 (-11.49 to 3.47)</strong></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 38.68; \chi^2 = 31.52$, df = 2 ($p < 0.00001$); $I^2 = 94\%$

Test for overall effect: $z = 1.05$ ($p = 0.29$)

**FIGURE 31** Meta-analysis SF-36 – general health perception.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Telephone</th>
<th>No telephone</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001112</td>
<td>31.22</td>
<td>22.68</td>
<td>35.4</td>
<td>26.12</td>
</tr>
<tr>
<td>Perkins 1998114</td>
<td>63.22</td>
<td>22.78</td>
<td>60.57</td>
<td>21.48</td>
</tr>
<tr>
<td>Weinberger 1996118</td>
<td>29.6</td>
<td>21.1</td>
<td>39.7</td>
<td>25</td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td><strong>2059</strong></td>
<td><strong>2202</strong></td>
<td><strong>100.0</strong></td>
<td><strong>-3.14 (-9.02 to 2.75)</strong></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 22.35; \chi^2 = 18.99$, df = 2 ($p < 0.0001$); $I^2 = 89\%$

Test for overall effect: $z = 1.04$ ($p = 0.30$)

**FIGURE 32** Meta-analysis SF-36 – vitality.
### Study or subgroup

<table>
<thead>
<tr>
<th></th>
<th>Telephone</th>
<th>No telephone</th>
<th></th>
<th>Mean difference</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001</td>
<td>53.84</td>
<td>58.76</td>
<td>37.7</td>
<td>−4.92 (−6.80 to −3.04)</td>
<td></td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>80.38</td>
<td>76.33</td>
<td>37.3</td>
<td>4.05 (1.77 to 6.33)</td>
<td></td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>58.9</td>
<td>67.5</td>
<td>24.9</td>
<td>−8.60 (−17.54 to 0.34)</td>
<td></td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>2059</td>
<td>2202</td>
<td>100.0</td>
<td>−2.49 (−9.98 to 5.00)</td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 37.74; \chi^2 = 37.91, \text{df} = 2 (p < 0.00001); I^2 = 95%$

Test for overall effect: $z = 0.65 \ (p = 0.51)$

---

**FIGURE 33** Meta-analysis SF-36 – mental health.

<table>
<thead>
<tr>
<th></th>
<th>Telephone</th>
<th>No telephone</th>
<th></th>
<th>Mean difference</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jones 2001</td>
<td>39.15</td>
<td>45.77</td>
<td>37.5</td>
<td>−6.62 (−8.59 to −4.65)</td>
<td></td>
</tr>
<tr>
<td>Perkins 1998</td>
<td>77.75</td>
<td>73.73</td>
<td>35.8</td>
<td>4.02 (0.49 to 7.55)</td>
<td></td>
</tr>
<tr>
<td>Weinberger 1996</td>
<td>40.1</td>
<td>45.5</td>
<td>26.7</td>
<td>−5.40 (−14.07 to 3.27)</td>
<td></td>
</tr>
<tr>
<td><strong>Total (95% CI)</strong></td>
<td>2059</td>
<td>2202</td>
<td>100.0</td>
<td>−2.49 (−10.62 to 5.64)</td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 44.86; \chi^2 = 26.71, \text{df} = 2 (p < 0.00001); I^2 = 93%$

Test for overall effect: $z = 0.60 \ (p = 0.55)$

---

**FIGURE 34** Meta-analysis SF-36 – bodily pain.
depression, hysteria, psychopathic deviation, masculinity–femininity, paranoia, psychasthenia, schizophrenia, hypomania and social introversion). The four remaining scales are included to safeguard against participants giving false results. The four validity scales are 'cannot say', lie, infrequency and defensiveness. The raw scores from each scale are hard to understand and are therefore transformed into standardised version of the score (T-score). Each standardised scale is scored on a range from 0 to 100 to aid interpretation.

Nine studies\(^{129–137}\) published between 1994 and 2003 used the MMPI. Not all of the studies reported all subscales. As for the SF-36, the impact of different modes of using the MMPI is assessed using weighted pooled measures of agreement for within-subject comparisons and random-effects meta-analysis for between-subject comparisons. Unlike the SF-36, however, the only mode comparison available for the MMPI was 'computer' versus 'not computer'.

Of the nine studies\(^{129–137}\) that reported the use of the MMPI, data were available from three studies\(^{129,131,134}\) for between-subject comparisons only, one study\(^{136}\) for within-subject comparisons only and three studies\(^{132,133,135}\) for both. Table 19 summaries the information available from each study.

### TABLE 19 Included studies using MMPI

<table>
<thead>
<tr>
<th>Paper</th>
<th>Country</th>
<th>Population</th>
<th>Design</th>
<th>Comp</th>
<th>Adm</th>
<th>Tel</th>
<th>Sens</th>
<th>Data availability</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Between-subject comparisons</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Biskin (1977)(^{129})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Randomised trial</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data available</td>
</tr>
<tr>
<td>Evan (1969)(^{130})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Randomised trial</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>No data</td>
</tr>
<tr>
<td>Hart (1985)(^{131})</td>
<td>USA</td>
<td>Male psychiatric referrals</td>
<td>Randomised trial</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data available for all scales other than the psychopathic deviant scale</td>
</tr>
<tr>
<td>Honaker (1988)(^{132})</td>
<td>USA</td>
<td>General population</td>
<td>Repeated measures</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken prior to crossover</td>
</tr>
<tr>
<td>Lambert (1987)(^{133})</td>
<td>USA</td>
<td>Substance abusers</td>
<td>Latin squares</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken prior to crossover</td>
</tr>
<tr>
<td>Locke (1995)(^{134})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Randomised trial</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data available only for the F scale(^{a})</td>
</tr>
<tr>
<td>White (1985)(^{135})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Crossover</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken prior to crossover</td>
</tr>
<tr>
<td><strong>Within-subject comparisons</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Honaker (1988)(^{132})</td>
<td>USA</td>
<td>General population</td>
<td>Repeated measures</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken combining order groups</td>
</tr>
<tr>
<td>Lambert (1987)(^{133})</td>
<td>USA</td>
<td>Substance abusers</td>
<td>Latin squares</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken combining order groups</td>
</tr>
<tr>
<td>Pinsoneault (1996)(^{136})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Randomised crossover</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken combining order groups</td>
</tr>
<tr>
<td>Shuldberg (1988)(^{137})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Crossover</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>No SDs</td>
</tr>
<tr>
<td>White (1985)(^{135})</td>
<td>USA</td>
<td>Psychology students</td>
<td>Crossover</td>
<td>y</td>
<td>n</td>
<td>n</td>
<td>y</td>
<td>Data taken combining order groups</td>
</tr>
</tbody>
</table>

Adm, administration; Comp, computerisation; n, no; Sens, sensory stimuli; Tel, telephone; y, yes.

\(^{a}\) See Appendix 7 for details of F scale.

Shaded cells indicate that these studies contributed to the analysis of that mode feature.
**Between-subject comparisons**

Six studies had data available that could contribute to the meta-analysis. Data taken from studies with a within-subject design have used data prior to any crossover. One study only had data available for one of the 14 scales. One study had data available for all scales other than ‘psychopathic deviant’. One study had no data that could be used for the meta-analysis.

**Within-subject comparisons**

Four studies had data available that could contribute to the within-subjects meta-analysis. One study, Lambert et al., had data available for all subscales other than the ‘cannot say’ scale. One study provided only mean scores, so could not contribute to the meta-analysis.

**Mode feature: computer**

All included studies that measured MMPI did so comparing ‘computer administered’ versus ‘not computer administered’. The results of the meta-analysis for each subscale of the MMPI can be viewed in Figures 35–48 (forest plots in order of magnitude of difference).

The ‘cannot say’ scale of the MMPI suggested higher scores when administered without a computer, with a mean difference of over seven points on the scale (see Figure 35). Although this could imply that participants view a computer terminal as a more private mode of data capture, and are, hence, less likely to leave a question blank than if they had to complete the MMPI with another form of data capture. It is more likely that the computer-completed measures did not allow for leaving items unanswered without justification. None of the clinical scales showed any significant differences (see Figures 39–48).

The combined results for the within-subjects studies are given in Table 20.

There were no significant differences between modes of administration for any of the subscales in the within-subject studies.

<table>
<thead>
<tr>
<th>Outcome</th>
<th>n</th>
<th>Mean difference</th>
<th>SD (diff.)</th>
<th>95% CI</th>
<th>95% limits of agreement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hypochondriasis</td>
<td>172</td>
<td>–0.7</td>
<td>5.2</td>
<td>–1.4 to 0.1</td>
<td>–10.9 to 9.5</td>
</tr>
<tr>
<td>Depression</td>
<td>172</td>
<td>–0.5</td>
<td>6.2</td>
<td>–1.4 to 0.5</td>
<td>–12.6 to 11.7</td>
</tr>
<tr>
<td>Hysteria</td>
<td>172</td>
<td>–0.5</td>
<td>5.7</td>
<td>–1.3 to 0.4</td>
<td>–11.6 to 10.6</td>
</tr>
<tr>
<td>Psychopathic deviation</td>
<td>172</td>
<td>–0.2</td>
<td>5.1</td>
<td>–1.0 to 0.6</td>
<td>–10.1 to 9.7</td>
</tr>
<tr>
<td>Masculinity–femininity</td>
<td>172</td>
<td>0.0</td>
<td>5.2</td>
<td>–0.7 to 0.8</td>
<td>–10.1 to 10.1</td>
</tr>
<tr>
<td>Paranoia</td>
<td>172</td>
<td>–0.7</td>
<td>4.9</td>
<td>–1.5 to 0.0</td>
<td>–10.3 to 8.9</td>
</tr>
<tr>
<td>Psychasthenia</td>
<td>172</td>
<td>–0.7</td>
<td>7.4</td>
<td>–1.8 to 0.4</td>
<td>–15.3 to 13.9</td>
</tr>
<tr>
<td>Schizophrenia</td>
<td>172</td>
<td>–0.8</td>
<td>10.0</td>
<td>–2.3 to 0.7</td>
<td>–20.4 to 18.7</td>
</tr>
<tr>
<td>Hypomania</td>
<td>172</td>
<td>–0.4</td>
<td>4.0</td>
<td>–1.0 to 0.2</td>
<td>–8.3 to 7.5</td>
</tr>
<tr>
<td>Social introversion</td>
<td>172</td>
<td>–0.5</td>
<td>6.5</td>
<td>–1.5 to 0.4</td>
<td>–13.3 to 12.2</td>
</tr>
<tr>
<td>Cannot say</td>
<td>97</td>
<td>–0.1</td>
<td>0.5</td>
<td>–0.2 to 0.0</td>
<td>–1.1 to 0.9</td>
</tr>
<tr>
<td>L</td>
<td>172</td>
<td>0.1</td>
<td>1.7</td>
<td>–0.2 to 0.4</td>
<td>–3.2 to 3.4</td>
</tr>
<tr>
<td>F</td>
<td>172</td>
<td>–0.5</td>
<td>5.9</td>
<td>–1.4 to 0.4</td>
<td>–12.1 to 11.0</td>
</tr>
<tr>
<td>K</td>
<td>172</td>
<td>–0.3</td>
<td>3.9</td>
<td>–0.9 to 0.3</td>
<td>–8.0 to 7.4</td>
</tr>
</tbody>
</table>

diff., difference.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Mean difference</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>15.75</td>
<td>18.43</td>
<td>37</td>
<td>1.89</td>
</tr>
<tr>
<td>White 1986</td>
<td>15.16</td>
<td>20.357</td>
<td>50</td>
<td>1.5</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>1.4</td>
<td>3.099</td>
<td>40</td>
<td>1.1</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>4.4</td>
<td>6.7</td>
<td>10</td>
<td>1.6</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>5.0</td>
<td>3.325</td>
<td>50</td>
<td>5.44</td>
</tr>
<tr>
<td>Total</td>
<td>137</td>
<td>145</td>
<td>100.0</td>
<td>7.23</td>
</tr>
<tr>
<td>Heterogeneity:</td>
<td>$\tau^2 = 0.29$, $\chi^2 = 36.29$, df = 3 ($p &lt; 0.00001$); $I^2 = 92%$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Test for overall effect:</td>
<td>$z = 2.04$ ($p = 0.04$)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**FIGURE 35** Meta-analysis MMPI – cannot say.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Mean difference</th>
<th>Mean difference</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>6.432</td>
<td>4.616</td>
<td>37</td>
<td>6.113</td>
</tr>
<tr>
<td>White 1986</td>
<td>5.02</td>
<td>3.325</td>
<td>50</td>
<td>5.44</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>55.9</td>
<td>8.415</td>
<td>40</td>
<td>54.65</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>12.19</td>
<td>7.91</td>
<td>38</td>
<td>11.45</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>63.8</td>
<td>13.7</td>
<td>10</td>
<td>67.9</td>
</tr>
<tr>
<td>Locke 1995</td>
<td>4</td>
<td>2.1</td>
<td>54</td>
<td>3.8</td>
</tr>
<tr>
<td>Total</td>
<td>229</td>
<td>290</td>
<td>100.0</td>
<td>0.12</td>
</tr>
<tr>
<td>Heterogeneity:</td>
<td>$\tau^2 = 0.00$, $\chi^2 = 1.80$, df = 5 ($p = 0.88$); $I^2 = 0%$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Test for overall effect:</td>
<td>$z = 0.41$ ($p = 0.68$)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**FIGURE 36** Meta-analysis MMPI – F.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Biskin 1977(^{129})</td>
<td>13.297</td>
<td>14.311</td>
<td>19.7</td>
<td>-1.01 (-3.39 to 1.36)</td>
</tr>
<tr>
<td>White 1986(^{135})</td>
<td>13.22</td>
<td>13.28</td>
<td>42.1</td>
<td>-0.06 (-1.69 to 1.57)</td>
</tr>
<tr>
<td>Honaker 1989(^{30})</td>
<td>54.15</td>
<td>54.6</td>
<td>7.1</td>
<td>-0.45 (-4.42 to 3.52)</td>
</tr>
<tr>
<td>Lambert 1987(^{133})</td>
<td>10.35</td>
<td>12.16</td>
<td>29.7</td>
<td>-1.81 (-3.75 to 0.13)</td>
</tr>
<tr>
<td>Hart 1985(^{31})</td>
<td>50.1</td>
<td>49.1</td>
<td>1.5</td>
<td>1.00 (-7.59 to 9.59)</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.78 (-1.83 to 0.28)</td>
</tr>
</tbody>
</table>

Heterogeneity: \(\tau^2 = 0.00; \chi^2 = 2.07, df = 4 (p = 0.72); I^2 = 0\%\)
Test for overall effect: \(z = 1.45 (p = 0.15)\)

FIGURE 37 Meta-analysis MMPI – K.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Biskin 1977(^{129})</td>
<td>2.72</td>
<td>2.622</td>
<td>28.4</td>
<td>0.10 (-0.85 to 1.05)</td>
</tr>
<tr>
<td>White 1986(^{135})</td>
<td>2.82</td>
<td>2.22</td>
<td>42.4</td>
<td>0.60 (-0.05 to 1.25)</td>
</tr>
<tr>
<td>Honaker 1989(^{30})</td>
<td>48</td>
<td>45.85</td>
<td>4.1</td>
<td>2.15 (-0.93 to 5.23)</td>
</tr>
<tr>
<td>Lambert 1987(^{133})</td>
<td>3.49</td>
<td>4.1</td>
<td>24.3</td>
<td>-0.61 (-1.68 to 0.46)</td>
</tr>
<tr>
<td>Hart 1985(^{31})</td>
<td>49.6</td>
<td>51.6</td>
<td>0.9</td>
<td>-2.00 (-8.87 to 4.87)</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>0.21 (0.44 to 0.85)</td>
</tr>
</tbody>
</table>

Heterogeneity: \(\tau^2 = 0.15; \chi^2 = 5.56, df = 4 (p = 0.23); I^2 = 28\%\)
Test for overall effect: \(z = 0.62 (p = 0.53)\)

FIGURE 38 Meta-analysis MMPI – L.
### Results

#### FIGURE 39 Meta-analysis MMPI – social introversion.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>28.892</td>
<td>10.661</td>
<td>37</td>
<td>26.4</td>
<td>10.532</td>
</tr>
<tr>
<td>White 1986</td>
<td>25.28</td>
<td>8.519</td>
<td>50</td>
<td>23.3</td>
<td>9.471</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>48.45</td>
<td>7.372</td>
<td>40</td>
<td>48.1</td>
<td>10.042</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>33.19</td>
<td>11.88</td>
<td>38</td>
<td>33.47</td>
<td>11.13</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>59.5</td>
<td>11</td>
<td>10</td>
<td>62.7</td>
<td>11.5</td>
</tr>
</tbody>
</table>

Total (95% CI) 175 182 100.0 −0.83 (−3.19 to 1.53)

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 1.73$, df = 4 ($p = 0.79$); $I^2 = 0$

Test for overall effect: $z = 1.02$ ($p = 0.31$)

#### FIGURE 40 Meta-analysis MMPI – masculinity–femininity.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>28.541</td>
<td>5.714</td>
<td>37</td>
<td>30.311</td>
<td>5.368</td>
</tr>
<tr>
<td>White 1986</td>
<td>32.72</td>
<td>8.261</td>
<td>50</td>
<td>30.18</td>
<td>7.041</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>54.85</td>
<td>13.653</td>
<td>40</td>
<td>53.9</td>
<td>12.055</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>23.95</td>
<td>4.53</td>
<td>38</td>
<td>25.18</td>
<td>4.8</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>62.1</td>
<td>8.8</td>
<td>10</td>
<td>69.7</td>
<td>6.3</td>
</tr>
</tbody>
</table>

Total (95% CI) 175 182 100.0 −0.83 (−3.19 to 1.53)

Heterogeneity: $\tau^2 = 3.87$; $\chi^2 = 9.83$, df = 4 ($p = 0.04$); $I^2 = 59$

Test for overall effect: $z = 0.69$ ($p = 0.49$)
### FIGURE 41 Meta-analysis MMPI – depression.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
<td>SD</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>18.243</td>
<td>5.118</td>
<td>37</td>
<td>19.311</td>
<td>5.888</td>
</tr>
<tr>
<td>White 1986</td>
<td>18.32</td>
<td>3.759</td>
<td>50</td>
<td>18.84</td>
<td>5.158</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>54.95</td>
<td>10.602</td>
<td>40</td>
<td>54.6</td>
<td>9.299</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>27.78</td>
<td>8.43</td>
<td>38</td>
<td>28.55</td>
<td>6.62</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>73.9</td>
<td>20.7</td>
<td>10</td>
<td>84.9</td>
<td>12.5</td>
</tr>
<tr>
<td>Total</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.71</td>
<td>(-1.96 to 0.55)</td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00; \chi^2 = 2.17, df = 4 (p = 0.70); I^2 = 0$

Test for overall effect: $z = 1.11 (p = 0.27)$

### FIGURE 42 Meta-analysis MMPI – paranoia.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
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<tr>
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<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
<td>SD</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>9.432</td>
<td>2.93</td>
<td>37</td>
<td>11.067</td>
<td>3.43</td>
</tr>
<tr>
<td>White 1986</td>
<td>9.64</td>
<td>2.85</td>
<td>50</td>
<td>9.6</td>
<td>2.999</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>56.7</td>
<td>10.493</td>
<td>40</td>
<td>55.65</td>
<td>8.25</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>13</td>
<td>5.05</td>
<td>38</td>
<td>13.9</td>
<td>5.11</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>61.8</td>
<td>12.1</td>
<td>10</td>
<td>69.4</td>
<td>12.6</td>
</tr>
<tr>
<td>Total</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.71</td>
<td>(-1.81 to 0.40)</td>
</tr>
<tr>
<td>Study or subgroup</td>
<td>Computer</td>
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<td>Mean difference IV, Random, 95% CI</td>
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</tr>
<tr>
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<td>-----------------------------------</td>
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</tr>
<tr>
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<td>Mean</td>
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</tr>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Weight (%)</td>
<td></td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>18.487</td>
<td>5.026</td>
<td>37</td>
<td>29.1</td>
<td>−0.80 (−2.85 to 1.25)</td>
</tr>
<tr>
<td>White 1986</td>
<td>18.88</td>
<td>4.237</td>
<td>50</td>
<td>39.6</td>
<td>0.14 (−1.61 to 1.89)</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>63.9</td>
<td>12.426</td>
<td>40</td>
<td>4.5</td>
<td>0.50 (−4.71 to 5.71)</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>21.35</td>
<td>5.28</td>
<td>38</td>
<td>25.7</td>
<td>−0.83 (−3.01 to 1.35)</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>64.5</td>
<td>8.5</td>
<td>10</td>
<td>1.1</td>
<td>−5.00 (−15.74 to 5.74)</td>
</tr>
<tr>
<td></td>
<td><strong>Total (95% CI)</strong></td>
<td></td>
<td>175</td>
<td><strong>182</strong></td>
<td>100.0</td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 1.48$, df = 4 ($p = 0.83$); $I^2 = 0$
Test for overall effect: $z = 0.75$ ($p = 0.45$)

FIGURE 43 Meta-analysis MMPI – hypomania.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
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<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Weight (%)</td>
</tr>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Weight (%)</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>5.054</td>
<td>3.274</td>
<td>37</td>
<td>31.8</td>
</tr>
<tr>
<td>White 1986</td>
<td>6.48</td>
<td>2.901</td>
<td>50</td>
<td>51.5</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>53.15</td>
<td>7.303</td>
<td>40</td>
<td>8.2</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>12.51</td>
<td>6.96</td>
<td>38</td>
<td>8.2</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>67.4</td>
<td>19.5</td>
<td>10</td>
<td>0.4</td>
</tr>
<tr>
<td></td>
<td><strong>Total (95% CI)</strong></td>
<td></td>
<td>175</td>
<td><strong>182</strong></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 2.17$, df = 4 ($p = 0.71$); $I^2 = 0$
Test for overall effect: $z = 0.51$ ($p = 0.61$)

FIGURE 44 Meta-analysis MMPI – hypochondriasis.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>18.865</td>
<td>3.772</td>
<td>37</td>
<td>21</td>
</tr>
<tr>
<td>White 1986</td>
<td>19.56</td>
<td>4.403</td>
<td>50</td>
<td>20.02</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>59.35</td>
<td>8.106</td>
<td>40</td>
<td>55.05</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>24.38</td>
<td>6.95</td>
<td>38</td>
<td>25.68</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>68.3</td>
<td>15.7</td>
<td>10</td>
<td>70.3</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.21</td>
</tr>
</tbody>
</table>

Heterogeneity: \( \tau^2 = 3.54; \chi^2 = 12.32, df = 4 (p = 0.02); I^2 = 68\%

Test for overall effect: \( z = 0.19 (p = 0.85) \)

**FIGURE 45** Meta-analysis MMPI – hysteria.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td>Mean</td>
</tr>
<tr>
<td>Biskin 1977</td>
<td>18.865</td>
<td>3.772</td>
<td>37</td>
<td>21</td>
</tr>
<tr>
<td>White 1986</td>
<td>19.56</td>
<td>4.403</td>
<td>50</td>
<td>20.02</td>
</tr>
<tr>
<td>Honaker 1989</td>
<td>59.35</td>
<td>8.106</td>
<td>40</td>
<td>55.05</td>
</tr>
<tr>
<td>Lambert 1987</td>
<td>24.38</td>
<td>6.95</td>
<td>38</td>
<td>25.68</td>
</tr>
<tr>
<td>Hart 1985</td>
<td>68.3</td>
<td>15.7</td>
<td>10</td>
<td>70.3</td>
</tr>
<tr>
<td>Total (95% CI)</td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.21</td>
</tr>
</tbody>
</table>

Heterogeneity: \( \tau^2 = 3.54; \chi^2 = 12.32, df = 4 (p = 0.02); I^2 = 68\%

Test for overall effect: \( z = 0.19 (p = 0.85) \)

**FIGURE 46** Meta-analysis MMPI – psychopathic deviation.
<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Biskin 1977\textsuperscript{129}</td>
<td>14.108</td>
<td>8.714</td>
<td>37</td>
<td></td>
<td></td>
</tr>
<tr>
<td>White 1986\textsuperscript{130}</td>
<td>15.86</td>
<td>6.443</td>
<td>50</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Honaker 1989\textsuperscript{132}</td>
<td>56.95</td>
<td>11.796</td>
<td>40</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lambert 1987\textsuperscript{133}</td>
<td>22.19</td>
<td>11.3</td>
<td>38</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hart 1985\textsuperscript{131}</td>
<td>70.2</td>
<td>17.3</td>
<td>10</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.13 (-1.98 to 1.72)</td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 3.28$, df = 4 ($p = 0.51$); $I^2 = 0$
Test for overall effect: $z = 0.14$ ($p = 0.89$)

**FIGURE 47** Meta-analysis MMPI – psychasthenia.

<table>
<thead>
<tr>
<th>Study or subgroup</th>
<th>Computer</th>
<th>No computer</th>
<th>Weight (%)</th>
<th>Mean difference IV, Random, 95% CI</th>
<th>Mean difference IV, Random, 95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Mean</td>
<td>SD</td>
<td>Total</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Biskin 1977\textsuperscript{129}</td>
<td>15.432</td>
<td>9.72</td>
<td>37</td>
<td></td>
<td></td>
</tr>
<tr>
<td>White 1986\textsuperscript{130}</td>
<td>13.86</td>
<td>6.491</td>
<td>50</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Honaker 1989\textsuperscript{132}</td>
<td>60.65</td>
<td>12.962</td>
<td>40</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lambert 1987\textsuperscript{133}</td>
<td>24.73</td>
<td>14.04</td>
<td>38</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hart 1985\textsuperscript{131}</td>
<td>75.6</td>
<td>19.5</td>
<td>10</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>175</td>
<td>182</td>
<td>100.0</td>
<td>-0.13 (-2.08 to 1.83)</td>
<td></td>
</tr>
</tbody>
</table>

Heterogeneity: $\tau^2 = 0.00$; $\chi^2 = 2.19$, df = 4 ($p = 0.70$); $I^2 = 0$
Test for overall effect: $z = 0.13$ ($p = 0.90$)

**FIGURE 48** Meta-analysis MMPI – schizophrenia.
Chapter 5
Discussion

The theoretical review has resulted in a change in focus from modes as discrete entities to be compared with a focus on mode features as factors relating to the way in which responses on subjective outcomes are constructed by responders. These primary features come from the model previously suggested by Tourangeau et al., with additional potential features identified. These have then been tested in the results from a comprehensive systematic review of mode comparison studies in terms of their impact on bias and precision.

The results of the review of mode comparison studies clearly show that the impact of mode features is on bias rather than precision. Therefore, in planning a new study, choice of mode and features is unlikely to have a great impact on sample size considerations, but may have an impact between single-mode studies on interpretability of values of scores and within mixed-mode studies on the ability to simply combine scores collected under different mode features. This lack of an impact on precision also suggests that different mode features do not lead to differing degrees of end aversion bias or floor/ceiling effects.

The mode feature with the greatest impact, in terms of both magnitude (size of effect) and significance (strength of evidence), was mode of administration (interviewer or self). The choice of sensory stimuli (audio or visual or both) had a smaller impact (about half that of mode of administration), but this was just significant. Neither computerisation nor telephone primary features were significant in the main models, although there was some suggestion of a potential difference that had decreased over time when interaction terms were tested. This fits with previous suggestions that mode features relating to technologies initially lead to differences predominantly due to unfamiliarity in the responder with the technology and that as technologies move into common usage these differences reduce.

Of the additional or secondary features tested, differences in mode of delivery reached significance in some of the models, but with a smaller magnitude than either administration or sensory stimuli. This feature was proposed as tapping into the perceived legitimacy of data collection in terms of how an outcome measure was delivered or presented to a potential responder.

In addition to the theoretically derived mode features, a small number of potential mediators were included in the model. Very limited information was available from studies on these and, therefore, the only two considered were date of publication in relation to the introduction of new technologies and the number of items in a scale to relate to part of the construct of cognitive burden. This latter factor was significant, with single-item scales showing a greater degree of bias than multi-item scales.

Overall, the primary analysis of the mode comparison studies identified in the systematic review provides consistent evidence for the impact of two of the four theoretical mode features having an impact on the absolute mean difference (bias), but not on precision. However, the magnitude of these effects, when considered on a percentage scale is not great. Further exploration into the two most frequently occurring scales within our review (SF-36 and MMPI) showed mixed results. The analysis for these was carried out for each mode feature individually and, therefore, the potential findings for telephone and computer features may well be due to the confounding...
nature of also having a difference in administration. However, the estimation of the pooled limits of agreement for the within-person studies emphasises the potential impact of mode effects if the purpose of measurement is to consider an individual rather than a group. At a group level the main analysis indicates that on average the bias is significant, but relatively small in ES-terms. However, if the measure is to be used in clinical practice, for example, then the reliability of the assessment of the individual becomes important and the limits of agreement show how variable this can be.

The SF-36 collects data on health status and, therefore, represents an example of potentially more sensitive data (an antecedent feature). This may, therefore, increase the chances of satisficing (for example) and the importance of ensuring privacy (impersonality) in data collection. In the first analysis, addressing the use of a computer, there is a clear mode feature effect present for the mental health domains of the SF-36, but not the physical health domains. If (in these studies) computers served to enhance impersonality, these results are consistent with the framework. A challenge in interpreting such results is a general lack of detail relevant to the psychological appraisal processes available in published reports.

**Strengths and weaknesses**

This was a broad and comprehensive systematic review in terms of breadth of the published literature covered and the independence of discipline. Innovative approaches to designing search strategies have been tested and implemented in order to produce a search strategy with high levels of specificity. Grey literature was not looked into, given the large volume of evidence produced from published papers and abstracts. The search strategy only covered the period up until 2004; however, a considerable number of studies were identified and contributed to the analysis. Future updates could take a more focused approach on new and emerging technologies.

The focus on the mode features rather than the crude modes is consistent with a theoretical basis to the analysis and also takes further the exploration of the strengths of proposed relationships from theoretical models. The review of theory and discussion within a health framework provides researchers with an understanding of the potential impact of these features when designing their study.

We were not able to test all the potential mode features, with anonymity in responding being one where few data were provided in papers. There was also limited information on potential mediating factors such as cognitive burden and sensitivity questions. Overall, presentation of information was highly variable, and some approach to standardising reports of these types of study would be recommended in the future if they are to inform researchers on the portability of measures across mode features.

The presented framework directed the design of the data extraction sheet for the systematic review. This was most important in relation to the mode features and antecedent features. For example, variables (levels) included in the data extraction form were administration (self or interviewer), sensory channel (auditory, visual or both) and computer-assisted data collection (yes, no, don’t know). Similarly, attempts were made to extract data related to the psychological appraisals. Thus, whether or not others were present at data collection and whether or not data collection ensured anonymity were both abstracted from studies. However, as expected, the availability of such data was limited in reviewed studies. Prospectively, a clear framework for conceptualising mode feature effects will be important for determining what data should be collected in empirical studies. Similarly, the analysis was guided by the framework, with key available variables from the framework included in the regression models. Hence, the initial regression model included all four mode features in the framework. Again, the lack of data about
framework features recorded in published work limited, to some extent, the scale of this analysis for some cases.

Conclusions

Researchers need to be aware of the different mode features that could have an impact on their results when selecting a mode of data collection for subjective outcomes. If researchers use a mixture of modes within their study (commonly a change in mode if there is poor or non-response) then consideration needs to be given to ameliorating potential biases consequent to this and controlling for them in analysis.

The potential does exist for there to be simple correction factors developed; however, these are likely to be measure specific. In analysis of current mixed-mode studies, researchers cannot just assume that results are comparable where a difference in administration or sensory stimuli exists and need to either undertake sensitivity analyses or formally control for mode in the analysis.

Recommendations for future research (in priority order)

There is growing recognition within health research of the need to consider measurement equivalence across modes. However, as evidenced in this review, there are already numerous studies considering a large number of outcome measures. However, these need to be reported in a standardised way to allow researchers to be able to make informed decisions about choice of mode with a particular outcome in a population. The development of reporting standards akin to PRISMA, STROBE (Strengthening the Reporting of Observational Studies in Epidemiology) or CONSORT (Consolidated Standards of Reporting Trials) for mode comparison studies is urgently needed and could build on the quality assessment tool developed here.

Prospective empirical studies need to be more theoretically informed (i.e. designed to measure and test theoretically relevant components) and to report accordingly. Greater attempts within such research are needed to understand whether or not the mode features are actually mediated in the way hypothesised.

Further mode comparison studies are required, but these need to be experimentally designed to manipulate mode features and directly assess the impact. This is preferable to more studies comparing two modes at a relatively pragmatic level without consideration of those features. Studies need to give consideration to evaluation and direct testing of the impact of some of the mediators of mode effects, as the lack of data presented in papers in this review limited our ability to analyse this component.

Further primary studies need to be undertaken to evaluate the impact of mode features over time. There was a suggestion across studies that this occurred for ‘new’ technologies for data collection (telephone and computer), but the ‘learning effect’ for any mode over time will be important to evaluate further in order to inform studies with long-term follow-up over multiple time points. The potential biasing impact of this ‘learning effect’ over time could be seen in single-mode studies as well as mixed-mode ones.

The focus of this review has been on measurement for research purposes and, therefore, has focused predominantly on the impact of mode features on estimated effects at a group level. However, the increasing use of subjective patient-reported outcomes in clinical practice means that considerable further work is required to consider measurement equivalence and reliability of assessment of individuals rather than groups.
Chapter 6

Dissemination

Publication


Oral presentations


2. Greene G, on behalf of the MODE ARTS Team. How does mode of survey administration affect the nature of the response provided? Some theoretical considerations. South West Society of Primary Care, Birmingham, UK, 2006.

3. Robling MR, on behalf of the MODE ARTS Team. Evaluating the impact of data collection mode upon response to subjective surveys: main results from the MODE ARTS systematic literature review. European Survey Research Association Biannual Conference, Prague, Czech Republic, 2007.


Poster presentations


2. Greene G, on behalf of the MODE ARTS Team. How does the modes of survey administration affect the response provided? All Wales Systematic Review Symposium, Cardiff, UK, 2006.

Projects/theses

1. Rhys Ivins. Analysis of the Minnesota Multiphasic Personality Inventory. Final Year Mathematics Undergraduate Project. Cardiff: Cardiff University; 2007

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Lesley Sander contributed to the design of the project and a co-applicant on the bid.

Naomi Cadbury contributed to the data extraction and quality assessment.

Kristina Bennert, Jo Crocker, Marie Ann Durand, Monika Hare, Dyfed Hughes and Sew Tien Yeo extracted data from the foreign-language papers.

Jonathon Gillard supervised the industrial placement student and advised on analysis.

Amanda Iles and Aude Espinasse undertook data entry.

Contribution of authors

KH led on the design and management of the study, conducted the analysis of the mode comparisons studies and wrote the report.

MR led on the theory review and contributed to the design, management, paper review, interpretation of analysis and final report.

DI contributed to the theory review and to the design, management, interpretation of analysis and final report.

DG contributed to the analysis and final report.

GG undertook abstract and paper review, data extraction, quality assessment and contributed to the final report.

RI undertook data extraction, quality assessment, and contributed to the analysis and final report.

IR contributed to the design, management, interpretation of analysis and final report.

AS undertook the design and evaluation of the search strategy, abstract and paper review, and contributed to the final report.

CS contributed to the design, management, paper review, interpretation of analysis and final report.

JW contributed to the interpretation of analysis and final report.

All authors reviewed the final version of the manuscript.
References


49. de Leeuw ED. *Data quality in mail, telephone and face to face surveys*. Amsterdam: TT-Publikaties; 1992.


92. Khan KS, Ter Riet G, Glanville J, Sowden AJ, Kleijnen J (editors). *Undertaking systematic reviews of research on effectiveness. CRD's guidance for carrying out or commissioning reviews*. CRD's guidance for carrying out or commissioning reviews.
2nd edn. CRD Report No. 4. York: NHS Centre for Reviews and Dissemination (CRD),

93. de Leeuw ED, Hox JJ, Snijkers G. The effect of computer-assisted interviewing on data

94. Downs SH, Black N. The feasibility of creating a checklist for the assessment of the
methodological quality both of randomised and non-randomised studies of health care

95. Kmet L, Lee R, Cook L. *Standard quality assessment criteria for evaluating primary research
papers from a variety of fields*. Edmonton, AB: Alberta Heritage Foundation for Medical
Research; 2004.

96. Group TSoRT. A proposal for structured reporting of randomised controlled trials. The

97. Anello C, Fleiss JL. Exploratory or analytic meta-analysis: should we distinguish between

98. van den Noorgate W, Onghena P. Multi-level meta-analysis: a comparison with traditional

1999;8:161–78.

100. Bland JM, Altman DG. Measuring agreement in method comparison studies. *Stat Methods


102. Williamson PR, Lancaster GA, Craig JV, Smuth RL. Meta-analysis of method comparison

103. Moher D, Liberati A, Tetzlaff J, Altman DG, Group TP. Preferred Reporting Items for
Systematic Reviews and Meta-analyses: The PRISMA statement. *BMJ* 2009;339:b2535. doi:
10.1136/bmj.b2535.

104. Bowling A, Bond M, Jenkinson C, Lamping DL. Short Form 36 (SF-36) Health Survey
questionnaire: which normative data should be used? Comparisons between the norms
provided by the Omnibus Survey in Britain, the Health Survey for England and the Oxford


research using standardised scales: can it be carried out over the telephone? *Psychol Med*

107. Allison T, Ahmad T, Brammah T, Symmons D, Urwin M. Can findings from postal
questionnaires be combined with interview results to improve the response rate among

108. Ooijen MV, Ivens UI, Johansen C, Skov T. Comparison of a self-administered questionnaire

Lawrence Erlbaum; 1988.


137. Schuldberg D. The MMPI is less sensitive to the automated testing format than it is to repeated testing: Item and scale effects. *Comput Hum Behav* 1988;4:285–98.


Appendix 1

Search strategy

TABLE 21 Databases searched

<table>
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<tr>
<th>Databases</th>
<th>Indexed: from-2004</th>
<th>Database provider</th>
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</thead>
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<td><strong>Health</strong></td>
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<td></td>
</tr>
<tr>
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<td>Ovid</td>
</tr>
<tr>
<td>BNI</td>
<td>1985</td>
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<td>Ovid</td>
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<td>Ovid</td>
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<td>1966</td>
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<td>Old MEDLINE</td>
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<td>Ovid</td>
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<td><strong>Evidence-based medicine</strong></td>
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<td>WoK</td>
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<td><strong>Hand-searching</strong></td>
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</tr>
<tr>
<td>ASA – Survey Research Methods</td>
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ACP, American College of Physicians; AMED, Allied and Complementary Medicine Database; ASA, American Statistical Association; ASSIA, Applied Social Sciences Index and Abstracts; BNI, British Nursing Index; CCRCT, Cochrane Central Register of Controlled Trials; CDSR, Cochrane Database of Systematic Reviews; CINAHL, Cumulative Index to Nursing and Allied Health Literature; CSA, CSA Illumina; DARE, Database of Abstracts of Reviews of Effects; EMBASE, Excerpta Medica Database; SCI, Science Citation Index; SSCI, Social Science Citation Index; WoK, Web of Knowledge; Ovid, Ovid Technologies.

Final search strategy

**Search string 0–1**

Finalised search strategy implemented in all databases allowing for changes in field codes and thesaurus terms.

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1. Computer$ ti, ab OR Mini Computer$ OR Mini-Computer$ OR Minicomputer$ OR Micro Computer$ OR Micro-Computer$ OR Microcomputer$ OR Multi Media OR Multi-Media OR Multimedia OR ACAPI OR CAPI OR CAI OR CACPI OR Touch Screen$ OR Touch-Screen$ OR Touchscreen$ OR Portable Computer$ OR Portable-Computer$ OR Portables computer$ OR PDA OR PDAs OR PDAs OR Personal Digital Assistant$ OR Personal-Digital-Assistant$ OR Personaldigitalassistant$ OR Personal-Digital Assistant$ OR Portable Digital-Assistant$ OR Pocket PCS OR Pocket-PCS OR Pockets pc$ OR Palm OR Psion OR Pocket Computer$ OR Pocket-Computer$ OR Lap Top$ OR Lap-Top$ OR Laptop$ OR Notebook$ OR Note Book$ OR Note-Book$ OR Pen Tablet$ OR Pen-Tablet$ OR Pentablet$ OR Virtual OR Interactive OR E mail$ OR E-mail$ OR Email$ OR Electronic Mail$ OR Electronic-Email$ OR Electronicmail$ OR Electronic Diar$ OR Electronic-Diar$ OR Electronicdiar$ OR HHC OR CAI OR ACASI OR PTC OR Palm Top OR Palm-Top OR Palmtop OR E-Diary OR Ediary OR Automated OR [Technology Assisted Thesaurus Terms]

2. World-Wide-Web OR World-Wide Web OR World Wide Web OR Worldwide Web OR WWW OR On Line OR Online OR On-line OR Internet$ OR Inter-Net$ OR Inter Net$ OR Intranet$ OR Intra-Net$ OR Intra Net$ OR Web Based OR Web-Based OR Webbased

3. Offline OR Off Line OR Off-Line OR Unplugged OR Un Plugged OR Un-Plugged


5. Facsimile OR Fax OR Telefax OR Telefacsimile

6. Telephone$ OR Cellular Phone OR Cellular-Phone$ OR Cellularphone Phone$ OR CATI OR CACI

7. Face to Face OR Facetoface OR Face-to-Face OR Interview$ OR Door to Door OR Door-to-Door OR Door to Door OR Door-to-Door OR Door to Door OR Curb Side OR Curbside OR Curbside OR Face-to Face OR Face to Face OR Person to Person OR Person-to-Person OR Person to Person OR Person to Person OR FTFI OR FTF OR F2F

8. Mode OR Modes OR Modal

9. Video$
Pocket-PCs OR Pocketpc$ OR Portable Computer$ OR Portable-Computer$ OR Portable Computer$ OR Postal OR Posted OR PPQ OR P&P OR Psion$ OR PTC OR Palm Top OR Palm-Top OR Palmtop OR Questionnaire$ OR SAQ OR Self Administ$ OR Self-Administ$ OR Selfadminist$ OR Self Answer$ OR Self-Answer$ OR Selfanswer$ OR Self Complet$ OR Self-Complet$ OR Selfcomplet$ OR Self Interview$ OR Self-Interview$ OR Selfinterview$ OR Self Disclosure OR Self Report$ OR Self-Report$ OR Selfreport$ OR Snail Mail OR Snail-Mail OR Snailmail OR Technology OR Telephone$ OR Touch Screen$ OR Touch-Screen$ OR Touchscreen$ OR Traditional OR Unplugged OR Un Plugged OR Un-Plugged OR Valid$ OR Video$ OR Virtual OR Web OR Webbased OR World-Wide-Web OR WWW

11. Alternat$ OR Blind$ OR Compar$ OR Concurrence OR Consist$ OR Contrast$ OR Control$ OR Cross Over OR Crossover OR Cross-Over OR Differ$ OR Error$ OR Evaluat$ OR Feasibility OR Group$ OR Mask$ OR Method.kw OR Methodolog$ OR Random$ OR Reliab$ OR Reproducibility of Results OR Sensitivity OR Specificity OR Survey OR Valid$ OR Versus$ OR Vs OR V’s

12. Administration$ OR Assessment$ OR Data Collect$ OR Diaries OR Diary OR Examination$ OR Interview$ OR Questionnaire$ OR Screen$ OR Self-report$ OR Survey$ OR Test$ OR [Comparative Thesaurus Terms]

13. 1 AND (OR/2 – 9)
14. 2 AND (OR/3 – 9)
15. 3 AND (OR/4 – 9)
16. 4 AND (OR/5 – 9)
17. 5 AND (OR/6 – 9)
18. 6 AND (OR/7 – 9)
19. 7 AND (OR/8 – 9)
20. 8 AND 9
21. OR/13 – 20
22. 10 AND 11 AND 12 AND 21
23. Limit 22 to Human
24. Limit 23 to yr = [Start Date – 2004]
Appendix 2

Data extraction sheets
Datasheet 1: full-paper initial screen

PAPER ID: 
Extracted by: Date of extraction: 

Does this paper compare 2 or more modes of data collection*? Y / N (if N, then STOP) 

Modes compared? 

Levels of reporting: 
Response rates Y / N Data quality Y / N 

Is the measurement of the same construct compared across different modes? Y / N (if N, then STOP) 

Does the comparison involve a diagnostic interview? Y / N 

<table>
<thead>
<tr>
<th>Measure</th>
<th>Construct</th>
<th>Subjective*, self-report? (=? = for discussion, judgment = N*)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Y / N / ?</td>
</tr>
<tr>
<td></td>
<td></td>
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<td>Y / N / ?</td>
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<td></td>
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<td>Y / N / ?</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Y / N / ?</td>
</tr>
</tbody>
</table>

WITHIN DESIGN 
- is the mode effect confounded by time of data collection* Y / N 

BETWEEN DESIGN 
- is the mode effect confounded by the sampling strategy* Y / N 

Notes: 

DECISION: IN OUT FOR DISCUSSION 

* SEE DEFINITIONS FOR CLARIFICATION
DEFINITIONS

MODE COMPARISON
A mode comparison study is one in which the same construct is measured (either with or without
the same tool administered) in two different modes, the scores are computed in the same way, and
the scores are (or can be) compared.

SUBJECTIVE
A subjective construct is one that is only accessible through an individual’s subjective self-report
(whether the self-report is recorded by the individual or by an interviewer or other person).

JUDGMENT
A study involves a judgment if the performance on the measure informs a judgment defined by an
external source e.g. a diagnosis, rather than the actual score derived from the measure.

WITHIN DESIGN
Confounds with the time of data collection relate to studies in which:-
  a) The use of two different collection methods that are not collecting data relating to the
     same time e.g. the use of a daily diary vs. a bi-weekly telephone interview

BETWEEN DESIGN
Confounds in the sampling strategy are
  a) when the sampling frame for groups are determined by different methods, e.g. door-to-
     door interviews within a small community (city block) vs. random digit dialling of a much
     larger community (city)
Datasheet 2: paper ID – paper information

**PAPER ID:**

<table>
<thead>
<tr>
<th>Extracted By:</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>Date Extracted:</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>Paper Name:</th>
</tr>
</thead>
</table>

**Source of Publication:**

- [ ] Health/health science (1)
- [ ] Psychology (2)
- [ ] Education (3)
- [ ] Social science (4)
- [ ] Business (5)
- [ ] Other (6)

**Exclude:**

- [ ] Yes (1)
- [ ] No (0)

**Exclude Reasons:**

<table>
<thead>
<tr>
<th>Country of Data Collection:</th>
</tr>
</thead>
</table>

**Language of Survey:**

- [ ] English (1)
- [ ] English (Assumed) (2)
- [ ] Other (3)

**Approached by:**

- [ ] University/Academic (1)
- [ ] Healthcare Trust/Hospital (2)
- [ ] Other Public Body (3)
- [ ] Provider/Insurance (4)
- [ ] Other Private Company (5)
- [ ] Charitable Body (6)
- [ ] Other (6)
- [ ] Don't Know (7)

**Design:**

- [ ] Within Groups (1)
- [ ] Between Groups (2)
- [ ] Both (3)
### Datasheet 3: sample and demographics data

**PAPER ID:**

<table>
<thead>
<tr>
<th>Sample and Demographics Data</th>
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<tbody>
<tr>
<td>Population:</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Site of Data Collection:</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Data Collection Team:</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Date of Data Collection:</td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>

**Time Frame:**

<table>
<thead>
<tr>
<th>Units:</th>
</tr>
</thead>
<tbody>
<tr>
<td>□ Hours (1)</td>
</tr>
<tr>
<td>□ Days (2)</td>
</tr>
<tr>
<td>□ Months (3)</td>
</tr>
<tr>
<td>□ Years (4)</td>
</tr>
</tbody>
</table>

**Sampling Strategy:**

<table>
<thead>
<tr>
<th>RDD (1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>□ Targeted (2)</td>
</tr>
<tr>
<td>□ Targeted - Clinic Lists (3)</td>
</tr>
<tr>
<td>□ Convenience (4)</td>
</tr>
<tr>
<td>□ Random (5)</td>
</tr>
<tr>
<td>□ Random Stratified (6)</td>
</tr>
<tr>
<td>□ Systematic (7)</td>
</tr>
<tr>
<td>□ Stratified (8)</td>
</tr>
</tbody>
</table>

**Target Time Gap (T1-T2):**

<table>
<thead>
<tr>
<th>Units:</th>
</tr>
</thead>
<tbody>
<tr>
<td>□ Hours (1)</td>
</tr>
<tr>
<td>□ Days (2)</td>
</tr>
<tr>
<td>□ Months (3)</td>
</tr>
<tr>
<td>□ Years (4)</td>
</tr>
</tbody>
</table>

**Justification of Time Gap:**

**Mean Achieved Time Gap:**

**SD Achieved Time Gap:**

**Range of Achieved Time:**

**Order Allocation:**

<table>
<thead>
<tr>
<th>All the Same (1)</th>
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<tbody>
<tr>
<td>□ Random (2)</td>
</tr>
<tr>
<td>□ Systematic (3)</td>
</tr>
<tr>
<td>□ Sampling (4)</td>
</tr>
<tr>
<td>□ Other (5)</td>
</tr>
</tbody>
</table>

**Group Allocation Other:**

**Population Description:**

**Personality Description:**
### PAPER ID:

No. of Modes Compared:

<table>
<thead>
<tr>
<th></th>
<th>N=Females</th>
<th>Age</th>
<th>SD Age</th>
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</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Ethnicity:

Educational Status:

SES:

Employment:

Notes:

Rewards:
- [ ] Yes (1)
- [ ] No (0)

Details:

Relevance:
- [ ] Yes (1)
- [ ] No (0)
- [ ] Not Sure (Details) (2)

Details:

Knowledge of Repeated Design:
- [ ] Yes (1)
- [ ] No (0)
- [ ] Not Sure (2)
Datasheet 4: mode description

PAPER ID:

Mode Description

Mode:
- Telephone Interview (1)
- VRE (2) / IVR (3)
- SAQ (4)
- DBM (5) / PDE (6) / CASI (7) / WS (8)
- TDE (9)
- PAPI (10)
- CAPI (11)
- ASAQ (12)
- VCASI (13)
- CATI (14)
- ACASI (15)

Method of Delivery:
- Telephone (Voice) (1)
- Telephone (Fax) (2)
- In Person (3)
- Mail (4)
- Email/Internet (5)

Computer Assisted Data Collection:
- Yes (1)
- No (2)
- Don't Know (3)

Administered by:
- Interviewer (1)
- Self/Respondent (2)

Sensory Channel:
- Auditory (1)
- Visual (2)
- Auditory & Visual (3)

Sensory Channel Notes:

Mode of Response:
- Oral (1)
- Written (2)
- Electronic (3)
- Other (4)

Mode of Response other:

Online vs. Off-line:
- Online (1)
- Off-line (2)

Presence of Others (interviewer):
- Yes (1)
- No (0)

Presence of Others (any):
- Yes (1)
- No (0)
- Don't Know (2)
### PAPER ID:

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<table>
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<tbody>
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</tbody>
</table>
Datasheet 5: measure description

PAPER ID:

Measure Description

Measure name:

☐ Immediate (1) ☐ Contemporary (2) ☐ Retrospective (3)

Time Frame:

Sub Construct:

Number of Items:

Lowest Value:

Highest Value:

Response Option Type:

☐ Likert-Like (1) ☐ VAS (2) ☐ Dichotomous (3) ☐ Categorical (normal) (4)
☐ Categorical (ordinal) (5) ☐ Other (6)

Response Option other:

Response Levels n=:

☐ Cms (1) ☐ Points (2) ☐ Events (3)

Cut off:

Construct Family:

☐ Health (1) ☐ Non-Health (2) ☐ Unknown (3)

Construct Family Unknown:

Construct Measure:

☐ Anxiety (1) ☐ Attitudes (2) ☐ Beliefs (3) ☐ Mental Health (4)
☐ Pain (5) ☐ Personality (6) ☐ Preference (7) ☐ QOL (8)
☐ Symptoms (9) ☐ Functional Health Status (10) ☐ Other (11)

Construct Other:
**PAPER ID:**

<table>
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<th>Subjective:</th>
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Number of Subjective Items:

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Notes:
## Datasheet 6: mode comparison data

### Mode Comparison Data

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<tr>
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<td>VRE (2) /IVR (3)</td>
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<tr>
<td>DBM (5) /PDE (6) /CASI (7) /WS (8)</td>
<td>TDE (9)</td>
</tr>
<tr>
<td>CAPI (11)</td>
<td>ASAQ (12)</td>
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<tbody>
<tr>
<td>Mode Item Order:</td>
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<tr>
<td>Fixed (All Item) (1)</td>
<td>Not known (4)</td>
</tr>
<tr>
<td>All items, adaptive order (2)</td>
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</tr>
<tr>
<td>All Adaptive (3)</td>
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</table>

<table>
<thead>
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<th>Pop</th>
<th>Mode:</th>
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<tr>
<td>Telephone Interview (1)</td>
<td>VRE (2) /IVR (3)</td>
</tr>
<tr>
<td>DBM (5) /PDE (6) /CASI (7) /WS (8)</td>
<td>TDE (9)</td>
</tr>
<tr>
<td>CAPI (11)</td>
<td>ASAQ (12)</td>
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</table>

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<th>Measure:</th>
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</thead>
<tbody>
<tr>
<td>Mode Item Order:</td>
<td></td>
</tr>
<tr>
<td>Fixed (All Item) (1)</td>
<td>Not known (4)</td>
</tr>
<tr>
<td>All items, adaptive order (2)</td>
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<tr>
<td>All Adaptive (3)</td>
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<table>
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<tr>
<th><strong>Duration (mean)</strong></th>
<th><strong>Duration (SD)</strong></th>
<th><strong>Range</strong></th>
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<tr>
<td>Mode 1</td>
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<tr>
<td>Mode 2</td>
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<th><strong>Time Frame</strong></th>
<th><strong>Baseline</strong></th>
<th><strong>T2=</strong></th>
<th><strong>T3=</strong></th>
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<tr>
<td><strong>N</strong></td>
<td>Mode1</td>
<td>Mode2</td>
<td>Mode1</td>
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<tr>
<td><strong>Mean</strong></td>
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<tr>
<td><strong>SD</strong></td>
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<tr>
<td><strong>Cronbach’s Alpha</strong></td>
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<tr>
<td><strong>Mean Difference</strong></td>
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<tr>
<td><strong>SD Difference</strong></td>
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<tr>
<td><strong>N Per Comparison</strong></td>
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<td><strong>Correlation</strong></td>
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<tr>
<td><strong>Correlation P-Value</strong></td>
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<tr>
<td><strong>Non-Specific P-Value</strong></td>
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<tr>
<td><strong>Difference Test</strong></td>
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<tr>
<td><strong>P-Value</strong></td>
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<tr>
<td><strong>Non-Specific P-Value</strong></td>
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PAPER ID:

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<th>Correlation Type:</th>
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<td>☑ Pearson's (1)</td>
<td>☑ Spearman's (2)</td>
<td>☑ Kendal (3)</td>
<td>☑ Limits of Agreement (4)</td>
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<tr>
<td>☑ ICC (5)</td>
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</table>

<table>
<thead>
<tr>
<th>Any Mode Related Differences:</th>
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</thead>
<tbody>
<tr>
<td>☑ Yes (1)</td>
<td>☑ No (Identical) (0)</td>
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</table>

Mode Related Differences Details:

Notes:

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<tr>
<th>B1</th>
<th>B 1/2</th>
<th>B 0</th>
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</thead>
<tbody>
<tr>
<td>A</td>
<td></td>
<td></td>
</tr>
<tr>
<td>A 1/2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>A 0</td>
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</table>

Notes:
### Datasheet 7: quality assessment

**MODE ARTS: Quality Assessment Tool**

<table>
<thead>
<tr>
<th>Criteria</th>
<th>Yes (2/good)</th>
<th>Partial (1/fair)</th>
<th>No (0/poor)</th>
<th>N/A</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Is the hypothesis/aim/objective of the study clearly &amp; sufficiently described?</td>
<td>Easily identified in introduction/method. Specifies: purpose, subjects/target population, and specific associations under investigation.</td>
<td>Vague/incomplete reporting or some info has to be gathered from parts of the paper other than intro/background/objective section.</td>
<td>Question or objective not reported/incomprehensible.</td>
<td></td>
</tr>
<tr>
<td>2. Are the measures clearly described?</td>
<td>Full description of measures including either a full appended version or a detailed description and examples of questions used</td>
<td>Some description of measure with no appended version or example of questions</td>
<td>Badly defined description of the measure (if no example please note source article if available)</td>
<td></td>
</tr>
<tr>
<td>3. Are the modes clearly described?</td>
<td>Full description of modes including the description of the way in which the measure is implemented in each mode</td>
<td>Some description of modes with no explicit description of implementation of measure.</td>
<td>Badly or no description of mode comparison</td>
<td></td>
</tr>
<tr>
<td>4. Is the main question(s) linked to a strong theoretical framework?</td>
<td>Hypothesis and objectives fully described within the context of a rigorous theoretical framework</td>
<td>Hypotheses derived loosely from theory with no explicit references to actual, only generalised theories or established concepts</td>
<td>Hypothesis mentioned with no reference to theory</td>
<td></td>
</tr>
<tr>
<td>5. Is the study design well described &amp; appropriate? (If study question not given, infer from conclusions).</td>
<td>Design easily identified and well described.</td>
<td>Design and/or study question not clearly described, or design only partially addresses study question.</td>
<td>Design does not answer study question or design is poorly described.</td>
<td></td>
</tr>
<tr>
<td>6. Are the characteristics of participants clearly described (e.g. age, SES ethnicity)?</td>
<td>Sufficient relevant demographic information. Reproducible criteria used to categorise participants clearly defined.</td>
<td>Poorly defined criteria or incomplete demographic information.</td>
<td>No baseline/demographic info provided.</td>
<td></td>
</tr>
<tr>
<td>7. Are the differences in selection across groups or conditions clearly described?</td>
<td>Described and appropriate. Inclusion/exclusion criteria described and defined.</td>
<td>Selection methods not completely described, but no obvious inappropriateness. Or selection strategy likely to introduce bias but not enough to seriously distort results.</td>
<td>No information/ inappropriate information provided or selection bias which likely distorts results.</td>
<td></td>
</tr>
<tr>
<td>8. Are the study sample representative of the intended population</td>
<td>A full description of the target population is given with the sample selected in a non-biased manner.</td>
<td>Sample selected from a known population however, selection strategy likely introduces bias but not enough to seriously distort results</td>
<td>Sample recruited from an unknown population in an opportunistic fashion</td>
<td></td>
</tr>
<tr>
<td></td>
<td>How were participants allocated to conditions?</td>
<td>If randomisation appropriate: Evidence of well randomised design with a description of the method used (e.g. random number tables, block design).</td>
<td>No randomisation mentioned but a stratified sampling method is utilised (i.e. may be that full randomisation may not be possible).</td>
<td>Random allocation not mentioned although it would have been feasible and appropriate (and possible done).</td>
</tr>
<tr>
<td>---</td>
<td>-----------------------------------------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td></td>
<td>Are population characteristics (if measured &amp; described) controlled for and adequately described?</td>
<td>Appropriate control at design/analysis stage or randomised study with comparable baseline characteristics.</td>
<td>Incomplete control/ description. Or not considered but unlikely to seriously influence results.</td>
<td>Not controlled for and likely to seriously influence results.</td>
</tr>
<tr>
<td></td>
<td>Was consideration given for data collected at different times (within groups)</td>
<td>A well described hypothetical reason why data was collected from participants at different time points or comparison with matched historical data set</td>
<td>Data was collected at different times due to specific opportunity</td>
<td>No explanation for data collection at different time points, either by chance</td>
</tr>
<tr>
<td></td>
<td>Are the groups adequately compared across</td>
<td>The same measure or mode adapted measures are applied to both groups with full description of procedure</td>
<td>No clear description of comparison across responder groups only that the same measure was utilised</td>
<td>No description of methods of comparison between groups or measure application</td>
</tr>
<tr>
<td></td>
<td>Have the characteristics of non-responders or participants lost to follow-up been described?</td>
<td>Losses adequately reported &amp; not likely to affect results, Or no responders or participants lost to follow up</td>
<td>Losses not well reported, but small &amp; not likely to affect results.</td>
<td>No information or large losses of responders and likely to affect results.</td>
</tr>
<tr>
<td></td>
<td>Are the main findings clearly described?</td>
<td>Simple outcome data (e.g. mean/proportions) reported for all major findings.</td>
<td>Incomplete or inappropriate descriptive statistics.</td>
<td>No/inadequate descriptive statistics</td>
</tr>
<tr>
<td></td>
<td>Are methods of analysis adequately described and appropriate?</td>
<td>Described and appropriate.</td>
<td>Not reported but probably appropriate or some tests appropriate, some not.</td>
<td>Methods not described and cannot be determined.</td>
</tr>
<tr>
<td></td>
<td>Are estimates of variance reported for the main results?</td>
<td>Appropriate estimates provided (SD/SE, confidence intervals).</td>
<td>Undefined or estimates provided for some but not all outcomes.</td>
<td>No information.</td>
</tr>
<tr>
<td></td>
<td>Does the explanation of the results lie within the theoretical framework identified in the introduction</td>
<td>Clear and coherent description of results discussed in relation to previous established theoretical framework</td>
<td>Findings related to generalised theory with no specific relation to specific theory</td>
<td>Findings discussed with no consideration to previously mentioned theory</td>
</tr>
<tr>
<td></td>
<td>Are the conclusions supported by the results?</td>
<td>All conclusions supported by data.</td>
<td>Some of the major conclusions are supported by the data; some are not. Or speculative interpretations are not indicated as such.</td>
<td>None/few of major conclusions supported by the data.</td>
</tr>
</tbody>
</table>
Appendix 3

Original funding proposal

Aim

To identify generalisable factors affecting responses to different modes of data collection from a systematic review of the literature.

Objectives

- To review all studies comparing two modes of administration for subjective outcomes and assess the impact of mode of administration on response.
- To provide an overview of the theoretical models of survey response and how they relate to health sciences research.
- To explore the impact of findings for key identified health-related measures.
- To create an accessible resource for health science researchers which will advise on the impact of the selection of different modes of data collection on response.

Outputs

- Generalisable guidance as to differences in the nature of response between modes of data collection.
- Overview of the theory of survey response in relation to measures used in health sciences research.
- Online resource for researchers designing studies.
- Provide workshops to relevant audiences to disseminate the results.

Background

Many studies in health sciences research rely on subjective outcome measures of some form or another. The increasing recognition of the importance of subject attitude to, and perceptions of, health and services provision has led to a rapid growth in such measures. Few clinical trials, even with interventions pharmacological or surgical in nature, would be run today without measuring patients’ QoL and the assessment of the acceptability of the intervention being trialled. Survey methodologies (in, for example, the business, marketing, social and political sciences) have an entire literature of their own, covering theory to practice, much of which has been slow to be recognised in the health arena. Few health-related outcome development papers indicate a theoretical approach to eliciting survey response.
Survey response and mode of data collection: psychological theories and survey techniques

The lack of an accepted theoretical basis for survey response was highlighted by Albaum, who noted the distinction between survey techniques and underlying psychological models. Four general theoretical frameworks considered to be particularly relevant to marketing research were reviewed: social exchange theory, cognitive dissonance theory, self-perception theory and theories of commitment and involvement. Albaum et al. surveyed the awareness and application of these theoretical models among business researchers across the world and found greatest adherence to theories of commitment and involvement. However, this theoretical review focused upon response decision rather than data quality or nature, although they did comment on the relative application of different models across varying data collection modalities.

Survey non-response and increasing concerns about maintaining adequate levels of response have led researchers to seek to categorise different forms of non-response. For example, Groves and Couper distinguish non-response due to non-contact, refusal to cooperate and inability to participate. The use of incentives to maintain response has, in turn, fostered theoretical development about how such inducements work which, for example, have focused upon economic theories of incentives through to models describing a broader consideration of social exchange. Comprehensive theories of survey involvement have also been introduced and tested empirically (for example Groves et al.).

Cognitive approaches to surveying

More recently, a paradigm shift has been described within survey methodology from a statistical model focused upon the consequences of surveying error to social scientific models exploring the causes of error. Attempts to develop such theories of (a) survey error, (b) decisions to participate and (c) response construction have been brought under the general banner of the Cognitive Aspects of Survey Methodology (CASM) movement. Understanding and reducing measurement error, rather than sampling error is at the forefront of this endeavour but Tourangeau notes how the statistical and social scientific approaches are complimentary rather than mutually exclusive. The impetus for recent theoretical developments is very much provided by technological innovation and diversity and a requirement to understand the relative impact of different data collection modes upon survey response.

Several information processing models describing how respondents answer questions have been proposed which share a common core of four basic stages: comprehension of the question; retrieval of information from autobiographical memory; use of heuristic and decision processes to estimate an answer; and response formulation. These models describe mostly sequential processing, apart from that proposed by Willis. The models have contributed to efforts to identify and resolve cognitive response problems in self-report questionnaires and thereby improve data quality in surveys through the use of evaluative and experimental techniques. Examples of the former include cognitive respondent interviews. The potential application of cognitive models and evaluative techniques to subjective self-report in areas such as HRQoL has recently been encouraged.

A good example of a sequential information processing model is provided by Tourangeau et al. Their model encompasses the four stages described above: (a) comprehension of the survey item; (b) retrieval of relevant information; (c) utilisation of information in making a judgement; and (d) formulating a response. For each stage, there are associated processes identified, which
a respondent may or not use when answering a question. Each stage and each process may be a source of response error. The theory is proposed for examining and understanding response to questions about events and behaviour as well as inherently subjective states such as attitudes.

The increase in options for survey data collection

As indicated above, there has been a substantial expansion in the modes of data elicitation and collection available to survey researchers over the last 30 years. In 1998, Tourangeau and Smith identified six methods that may be employed for face-to-face interviews including paper-and-pencil personal interviews (PAPIs), paper-and-pencil self-administered questionnaires (SAQs), Walkman-administered questionnaires (audio-SAQs), computer-assisted personal interviews (CAPIs), computer-assisted self-administered interviews (CASIs) and audio computer-assisted self-administered interviews (ACASIs). Subsequently, Tourangeau et al. delineated 13 different modes of survey data collection (including remote data collection methods such as telephone, mail, e-mail and the internet), which they considered differed in terms of five characteristics: how respondents were contacted; the presentational medium (e.g. paper or electronic); method of administration (via interviewer or self-administered); sensory input channel used; and response mode.

Applying cognitive models to survey response modality

Psychological models of survey response have been applied to the issue of data collection mode. Tourangeau and Smith' proposed three characteristics of the data collection mode that may be affecting response; computerisation, whether a survey schedule is self- or interviewer administered, and whether survey items are read by or to the respondent. A fourth characteristic (the use of telephone) was included in a later formulation of this model. Three psychological variables are considered to mediate the impact of data collection mode; degree of privacy permitted (subsequently amended to ‘impersonality’), level of cognitive burden imposed and the sense of legitimacy engendered by the approach. The model hypothesises the effect of the mediating variables upon levels of reporting, accuracy, reliability and rate of missing data.

The model has still to be systematically evaluated although some evidence is available. For example, an important consideration has proven to be survey item sensitivity which may serve to emphasise differences between data collection modes (e.g. self-administration vs interviewer). Approval from the interviewer would appear to be the salient influence and may lead to either under- or over-reporting of behaviour depending upon its social acceptability. The level of privacy or degree of impersonality afforded by the data collection mode will thus differentially influence the impact of this tendency. While the studies non-systematically reviewed by Tourangeau and Smith involve behavioural self-report (some of which may be externally validated, e.g. alcohol consumption), other non-observable attitudes may be equally susceptible to such influences (e.g. social stereotyping, racial attitudes, etc.).

Variations even within the same mode of data collection further complicate evaluation. For example, Honaker describes computer administered versions of the MMPI, which differ in terms of type of computer being used, different computer–user interfaces with inconsistent item presentation and response formats. Therefore, results from one computerised version of a test cannot be easily generalised to other versions. Other variables that could mediate the effect of different modes of data collection have also been considered, including the overall pace of the interview, the order of survey item processing and role of different mental models employed.
by respondents. Although the latter in particular is rarely assessed, it has been considered a potentially significant mediator of response behaviour.8

Alternative cognitive approaches include work on optimising and satisficing, concepts described as two ends of a continuum of thoroughness of the response process.39 A respondent may proceed through each cognitive step less diligently when providing a question response or they may omit the middle two steps completely (i.e. retrieval and judgement) – examples of weak and strong satisficing, respectively. In either situation, a variety of decision heuristics may be utilised by the respondent to provide a satisfactory answer. The theory has been used to explain a variety of phenomenon observed in surveys, for example, response order effects (recency and primacy), which emphasise the role of scale design and mode of administration.

Holbrook et al.42 reviewed survey satisficing theory and another hypothetical source of measurement error, social desirability bias across telephone and face-to-face interviews. The probability of satisficing is a function of respondent ability, respondent motivation and task difficulty. Situational factors such as level of non-verbal communication, interview pace and respondent multitasking, which differ between modes interact with respondent disposition to affect response quality. Social desirability bias whereby respondents intentionally misrepresent themselves in their survey responses may differentially affect data collected via different modes. This could stem from differences in social distance, rapport and trust. Holbrook et al.42 found evidence that suggested that telephone interviews increased satisficing and social desirability response bias compared to face-to-face interviews. Also highlighted was the potential interaction of factors such as educational level.

The challenge for health sciences research

As described above, the first characteristic underlying the different modes of data collection considered by Tourangeau was method of contact.8 Work assessing the impact of an integrated process of respondent approach, consent and data collection has addressed bias due to selective non-ascertainment (i.e. the exclusion of particular subgroups). This may be clearly identifiable subgroups in terms of people without telephones or computers (for telephone or internet approaches), or less clearly identifiable subgroups, i.e. those with lower levels of literacy or the elderly (for paper-based approaches). There is also considerable work on improving response rates and the biases induced by certain subgroups being less likely to consent to take part in a survey.

Furthermore an important question in Health Sciences Research is the use of data collection methodologies within prospective studies, where patients have already been recruited via another approach. This could be within a clinic or other health service setting rather than the survey instrument being the method of approach as well as data collection. Edwards et al. have recently reviewed the literature (both health and non-health) to identify randomised trials of methods of improving response rates to postal questionnaires. Another recent review in health-related research1 has focused on the completeness of data collection and patterns of missing data, as well as response rates.

Indeed guidance is needed not just in terms of which is the ‘best’ method to use and most appropriate theoretical model of response, but also the possible effects of combining data collected via different modes as there is an increasing need for multimethod follow-up to capture all of the sample of interest. For example, a commonly observed multimethod approach is when a second mode of data collection is used when the first has been unsuccessful (e.g. using telephone interview when there has been no response to a postal approach11). Criteria for judging
equivalence of two approaches is therefore required. Honaker uses the concepts of psychometric equivalence and experiential equivalence. The former describes when the two forms produce results with equal mean scores, identical distribution and ranking of scores and agreement in how scores correlate with other variables. The latter deals with how two forms may differ in how they effect the psychometric and non-psychometric components of the response task.

In order to inform health services research, guidance is needed, which quantifies the differences between modes of data collection and indicates which factors are associated with the magnitude of this difference. These could be contextual based in terms of where the participant is when the information is completed (e.g. health setting, own home, work), content based in terms of questionnaire topic (e.g. attitudes to sexual behaviour) or population based (e.g. elderly). Previous work has shown moderate reliability between SAQ and interview on health problems in an elderly population post transurethral resection of the prostate. However, there was a consistent tendency for the SAQ to underestimate a patient's health problems compared with interview. The factors identified by Tourangeau also need to be tested across a wide range of modes and studies.

**Defining subjective outcomes in health sciences research**

Of particular interest in HSR is the collection of data which cannot be validated objectively. This results in a situation where there is no 'gold standard' with which to compare results to and therefore care needs to be taken as to the presumption of the 'correctness' of responses. This incorporates many types of outcome which are of key interest to health researchers, such as attitudes, intentions to behave and beliefs about illness. This type of outcome can be classified as evaluation-based, where the subjective perspective of the individual is an intrinsic component of the construct being measured. These can be distinguished from performance- and perception-based measures using the following example (from Schwartz and Rapkin):

- **Performance**  Timed walk up flight of stairs.
- **Perception**  How often do you walk up a flight of stairs?
- **Evaluation**  How difficult is it to walk up a flight of stairs?

The involvement of proxy raters in the assessment process for certain groups, particularly in health, is relatively common. For certain patient groups self-report may be difficult and another person is chosen to report on their behalf. All of the modes and much of the theoretical basis of response described above can be used to collect data about an individual via a proxy. This proxy may be a relative (such as a parent or spouse) or someone responding in a professional capacity, such as a health professional. The focus on an evaluation-based framework for outcome measures would lead to this type of measure being included when the comparison is of different methods of data collection within an individual (i.e. incorporating the same individual's subjective perspective). However, the subjective nature of evaluation-based outcomes which involve judgement using idiosyncratic criteria would lead to studies that compare proxy-reporting to self-reporting being excluded from this review.

**Review: direct comparisons of data collection modes (health and non-health-related outcomes)**

**Methodology**

*Overview:* an extensive search of both health and non-health literature will be conducted to identify studies which compare two or more modes of data collection on subjective measurement on the same scale.
Outcomes measures: evaluation-based measures such as attitude, satisfaction, belief, intention to behave, QoL constructs such as anxiety, pain, vitality (not physical functioning).

Studies: will need to have compared two or more modes of data collection in terms of the responses given. Studies purely considering response rates, data recording errors or costs will not be included. These studies will be identified by the search strategy and the reviews to date have limited themselves to postal and other self-completed surveys. Although it is not covered by this application due to cost limitations and is not specific to the remit of brief, this gives the opportunity to provide a database that can be analysed separately. Response rates and costs of using proxy raters and the impact of the use of information technology in either interviewer assisted or self-completed modes are valid questions still to be answered.

Topics: studies in any topic area, both health and non-health, will be included. There is considerable literature on the impact of different response options on outcome, therefore this review will be restricted to studies where the sole purpose of a different response scale is to accommodate the data collection mode. An example of this would be where a postal questionnaire uses a visual analogue scale, whilst a telephone interview would have to replace this with an ordinal one. This will be controlled for in the analysis.

Search strategy

McColl et al. started their review in 1975 with the justification that this was the decade in which several seminal works on surveys were published and the interest in survey methodology took off. However, Edwards' review on response rates identified a number of randomised trials of methods of increasing response rates to postal surveys published prior to this. Therefore, we intend to search electronic databases from the dates they are available. Box 1 gives the electronic databases used in previous systematic reviews of response rates (Edwards) and design issues (McColl) plus additional databases felt to be of relevance.

In addition to the above databases, the National Research register will be searched for ongoing relevant studies. Certain non-indexed highly relevant collections will be hand searched (e.g. the proceedings of the Survey Research Methods Section of the American Statistical Association). All included papers will have their reference lists searched for relevant papers, using a pearl-growing approach.

Negative publication bias is unlikely to be operating for this type of study, i.e. whether two methods are shown to be the same or different is unlikely to affect the chances of a study being published. Therefore, the search will be limited to published studies, so databases covering grey literature such as Index to Theses, Dissertation Abstracts and SIGLE will not be searched unless there is a lack of evidence for any particular mode of data collection.

The search strategy will focus on data collection mode with additional filters for identifying comparative studies. A matrix approach will be used to reduce the number of studies that report data using only a single mode of administration which are identified. This approach means that we will be searching for studies which contain any two of the following sets of terms:

- Question$ or paper or postal or mail
- Telephon$
- Computer$
- Interview$

A scoping search on MEDLINE from 1996 to 2004 gives the following number of hits (Table 1).
These will be combined with appropriate words for each database to focus on studies of reporting validation. In MEDLINE the Mesh term ‘Reproducibility of Results’ will be used. This reduces the number of hits in the above scope and produces a far more sensitive search (Table 2).

All searches will be limited to studies of humans. Non-English studies will be identified but only included where there is a lack of evidence in the comparison of any two particular modes.

All identified titles and abstracts will be downloaded into a Reference Manager database, duplicates removed and then titles and abstracts independently reviewed by two reviewers to assess eligibility for inclusion. Studies which either or both reviewers consider eligible will be retrieved in full. An assessment of chance corrected agreement (Kappa) will be made after every 100 abstracts reviewed as a form of quality control on the process. Full papers will again be reviewed for eligibility and data extracted by two reviewers. Additional searches will be made for the ten authors with the highest number of hits.
Eligibility criteria

Include studies:
- using two or more different modes of data collection used
- measuring an evaluation-based assessment
- where both modes of data collection are applied to the same measurement scale
- that have a comparative element either at an individual or group level.

Exclude studies:
- comparing proxy to self-completion
- focusing solely on response or error rates and cost of administration.

Level of comparative analysis

Individual level comparative studies will consist of individuals being exposed to both modes of data collection and their results being compared in a paired analysis. The highest level of evidence would come from those that randomised each individual as to the order in which the data collection modes were used. This type of study design is essentially a cross over trial, allowing for assessment of carry-over effect (recall bias). Additional quality criteria would be consideration and justification of the impact of the time lapse between approaches and the impact of participant recall and stability of the construct being measured.

Group level comparative studies would involve individuals being randomised (or quasi-randomised) to one of the modes of collection to be compared. Analysis would then be at a group level. Consideration would need to be given within the study to the level of balance that the randomisation/quasi-randomisation had achieved on other factors associated with the outcome.

Data extraction

A data extraction sheet will be developed and piloted covering standard quality markers for the reporting of studies, along with factors specific to individual and group level comparisons. A training set of papers \((n = 25)\) will be critically appraised and data extracted by two researchers. After this each paper will have data extracted by a single researcher except where difficulties arise. The extraction of statistical data and statistical modelling will be guided by Dr Hood and Prof. Russell.

Data could be reported in one of the following ways:
- means/mean differences (or proportions) with SE (e.g. Krysan et al.\(^{149}\))
percentage agreement or Kappa statistics (e.g. Doll et al.\textsuperscript{147})
variances or reliability coefficient.

All studies would be rated on their quality of reporting in terms of response rates/loss to follow-up, details of their follow-up procedures for non-response and patterns of missing data. Additional variables will rate the difference between the two methods being tested in terms of development and validation and the intensity of the follow-up process. This will include whether the design was theoretically based. With different modes of data collection, identical follow-up procedures are unlikely to be appropriate; however, they should be equivalent in terms of intensity. This leads to a measurement of quality of study that is based on the degree of similarity between the two approaches. This will be rated in terms of development/validation and intensity of follow-up (same/moderately different/very different). A quality scoring system based on this will be developed and controlled for in the analysis.

Key data defining how respondents were contacted; the presentational medium (e.g. paper or electronic); method of administration (via interviewer or self-administered); sensory input channel used (audio and/or visual); and response mode (verbal or manual) will be identified for each mode within each comparison. Where possible, the content, context and population will be categorised.

**Analysis**

Information from the data extraction sheet will be entered into SPSS for preliminary analysis. The studies should provide information on overall means/mean differences (for group/individual studies) and standard errors. These will be analysed using meta-regression to explore differences by mode of data collection and other variables of interest such as context, content and population. The dependent variable in these analyses will be the standardised difference between the two modes. The modes will be labelled according to the categories identified from theoretical cognitive models.\textsuperscript{8} This would involve modes of data collection categorised according to differences in the presentational medium, method of administration, sensory input and response mode. The impact of levels of computerisation will also be assessed. Where more than one outcome per study is of interest, a two-level model will be fitted (using MlWin) to allow for correlations between outcomes within a study. Assessment will be made whether a fixed or random effect fits best for each factor.

Possible moderating factors will be assessed, covering:

- Administration factors, such as intensity of follow-up.
- Population factors, such as age, social class, educational level and disease group.
- Scale-specific factors, such as number of items, response options, time taken and the theoretical basis for its development. A key variable to explore will be whether the scale is health related or not.

Certain modes of data collection may not be represented in enough of the identified studies to be included in the analysis of moderating effects. This analysis will enable us to ascertain the degree to which generalisable conclusions can be drawn across topic areas and populations.

Other factors that will be explored, provided enough studies are identified, are whether the magnitude of the differences between modes is affected by the number of items to be completed and the time taken. Certainly the degree of recall bias in individual studies may be affected by the number of items being completed.
Individual and group studies will be analysed separately. Sensitivity analysis will explore the impact on the conclusions drawn of weighting the regression by quality and sample size.

In order to show psychometric equivalence it is not enough for the mean differences to be close to zero, the distributional properties must also be the same. Therefore where possible comparison of the variances for the different modes of data collection will also be analysed. Again, group and individual studies will need to be analysed separately. This analysis will use the ratio of the variances for each mode from a study and explore whether particular modes of data collection lead to greater variability in response.

**Bringing the results into the health domain**

A key question in health sciences research is how generalisable the lessons learnt in other disciplines such as sociology and psychology are to the health field. Certain subjective constructs of interest in these other disciplines are more clearly related to outcomes we wish to measure, although whether the cognitive processes involved in responding are content specific remains to be shown. Therefore we propose to undertake two additional pieces of work.

**Review of theory**

Much of the theory of survey response is published in the survey methodology literature. Since an essential part of understanding the results from the review is to link it to theory, we will undertake a review of the psychological models which can explain/predict individual response to differing modes of data collection. This will be drawn together and interpreted for the health domain. This will be used to provide guidance for researchers developing new measures – for example, on particular validation assessments needed for different modes of data collection. More generally, it can also help guide good practice in the development, design and application of health outcome measures.

**Additional review/overview**

The systematic review above is limited to studies which have directly compared different modes of data collection. However, there is still a question whether this type of study generalises to those using a single mode of data collection. The focus on comparing modes of administration may make the studies ‘idealised’ to certain degree with the typical focus being on recruitment, retention and compliance issues rather than on the construct being measured per se. The presence of participant recall bias may under estimate differences between modes. There is therefore a value in considering whether similar patterns of differences exist in studies which use a single mode of administration to the comparative studies included in the review above.

Only a small number of studies within the health field have directly compared two or more modes of data collection. In contrast, a very large number of studies have each used a single mode of data collection. These single-mode studies can be used to assess the generalisability of the results of the review. The review will identify direct comparisons of generic health-related measures such as SF-36\(^1\text{04,113,114}\) and condition specific measures.\(^1\text{50}\)

In order to address this issue of external validity, we will review studies that have used one generic instrument, the SF-36, and up to three condition specific instruments identified during the search for direct comparisons. For these outcomes studies will be identified which have
administered (singly) the modes of administration which were directly compared. For the
generic measure (SF-36) most of the alternatives (telephone, interview) will have been compared
to paper-based questionnaires. In this case we will identify a sample of paper-based studies in
the same patient grouping as other modes have been used in. These will be analysed to explore
whether the magnitude of the differences between the measures (controlling for differences in
study design and population) shown in the direct comparison studies is born through to studies
using an individual mode.

Outputs

The results of this systematic review will show the magnitude of differences between different
modes of data collection and how this is affected by moderator variables such as context and
population. It will also explore further the theoretical framework proposed by Tourangeau.
Practical outputs could take the form of actual correction factors for modes with have been well
studied (and factors are shown to be generalisable) and more general guidance for the less well
studies ones.

The results of the review will be evaluated alongside the cognitive models proposed in the theory
of survey response. A particular focus will be on how the models related to measures used in
health. The additional review of individual measures will related the findings of from the general
review directly to the types of measures of interest in health.

Dissemination

A key component of the dissemination strategy is to provide an online resource for health
services researchers. This will include a database summarising all of the direct comparison
studies so that they can be easily searched and identified. However, a key component of this will
be where possible to indicate quantitatively how different modes will impact on the results during
the planning stage of a study. It is also hoped that this initial resource will be contributed to by
research teams using novel modes of data collection to provide an ongoing resource which is
both used by and contributed to by the whole research community. The ongoing maintenance of
such a resource would become part of the Centre for Health Science Research.

In addition to this a workshop will be held to discuss the theoretical perspectives in survey
response and how they relate to health. We will also look to target workshops at key conferences
such as the International Society for Quality of Life Research (ISOQOL) and the Society
for Social Medicine (SSM). In addition to this we will also offer workshops/seminars to key
organisations such as the Royal Colleges and the Royal Statistical Society. The components of the
review will be written up into a report and as peer reviewed publications in mainstream journals.

Management structure

The co-applicants will form a management team which will meet monthly by audio conference
and face-to-face once a quarter, starting with a face-to-face meeting. Members of this team have
worked across Wales (and the rest of the UK) using this combination of audio and face-to-face
meetings in the past successfully. Dr Kerry Hood, Mike Robling, Lesley Sander and the RA will
meet formally on a weekly basis between management meetings. Dr Kerry Hood will lead the
meetings, manage the project day to day and have line management responsibility for the two
employed staff. The review of theoretical models will be managed by Mike Robling and Dr David Ingledew. Prof Ian Russell will work with Dr Hood on the statistical modelling.

**Justification of resources and time frame**

This systematic review is across numerous electronic databases and from scoping searches is likely to identify a large number of studies. Therefore it is proposed to employ two members of staff for a year each. Lesley Sander is an experienced information scientist who is available to start immediately. She will initiate the project whilst we are appointing the RA. Resources are requested for PCs for both of these members of staff and an additional copy of Reference Manager (plus manual) for the RA. We estimate needing £2000 for inter-library loans. This cost has been kept down due to the fact that Cardiff University has extensive libraries and e-journals and therefore an amount is requested for photocopying and printing. The Proceedings of the Survey Methods Section of the American Statistical Association are available on CD which has been costed in along with the manual and the electronic bibliography for SF-36. Consultancy time of 12 days for input on the review of theory by Dr David Ingledew (@£500 per day) and 4 days for statistical modelling input from Prof. Ian Russell (@£1000 per day) has been costed.

Costs for travel and telephone for management meetings has been included for six face-to-face meetings (one initial followed by once a quarter) and three audio conferences per quarter.

We are planning the development and design of the web site with a company who undertake much work in the academic health field (waters-design) and have recently developed the website for the new Swansea Clinical School. An approximate costing for this has been put at £8000. In order to ensure dissemination via workshops, a conference budget of £3500 has been requested.

**Timetable**

<table>
<thead>
<tr>
<th>Month</th>
<th>Tasks</th>
</tr>
</thead>
</table>
| 0–3   | Refine search strategy  
|       | Draught data extraction sheet  
|       | Appoint RA  
|       | Run searches and remove duplicates  
| 3–6   | Assess abstracts for inclusion  
|       | Retrieve full papers and assess for inclusion  
|       | Pilot data extraction sheet  
|       | Identify specific health-related scales for single-mode studies and search  
| 6–9   | Extract data from included papers  
|       | Retrieve papers on single mode  
|       | Identify and retrieve theoretical papers  
| 9–12  | Enter extracted data  
|       | Analyse direct comparisons  
|       | Extract data for single-mode studies  
|       | Synthesise the theoretical papers  
| 12–15 | Analyse single-mode studies  
|       | Write up report and papers  
|       | Design web page  
|       | Conduct workshops |
# Appendix 4

The Preferred Reporting Items for Systematic Reviews and Meta-Analyses checklist

<table>
<thead>
<tr>
<th>Section/topic</th>
<th>#</th>
<th>Checklist Item</th>
<th>Reported on page #</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Title</strong></td>
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</tr>
<tr>
<td>Title</td>
<td>1</td>
<td>Identify the report as a systematic review, meta-analysis or both</td>
<td>i</td>
</tr>
<tr>
<td><strong>Abstract</strong></td>
<td></td>
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</tr>
<tr>
<td>Structured summary</td>
<td>2</td>
<td>Provide a structured summary including, as applicable: background; objectives; data sources; study eligibility criteria, participants; and interventions; study appraisal and synthesis methods; results; limitations; conclusions and implications of key findings; systematic review registration number</td>
<td>xi–xv</td>
</tr>
<tr>
<td><strong>Introduction</strong></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Rationale</td>
<td>3</td>
<td>Describe the rationale for the review in the context of what is already known</td>
<td>1–3</td>
</tr>
<tr>
<td>Objectives</td>
<td>4</td>
<td>Provide an explicit statement of questions being addressed with reference to participants, interventions, comparisons, outcomes and study design (PICOS)</td>
<td>3</td>
</tr>
<tr>
<td><strong>Methods</strong></td>
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<tr>
<td>Protocol and registration</td>
<td>5</td>
<td>Indicate if a review protocol exists, if and where it can be accessed (e.g. web address), and, if available, provide registration information including registration number</td>
<td>NA – original funding proposal pp. 103–114</td>
</tr>
<tr>
<td>Eligibility criteria</td>
<td>6</td>
<td>Specify study characteristics (e.g. PICOS, length of follow-up) and report characteristics (e.g. years considered, language, publication status) used as criteria for eligibility, giving rationale</td>
<td>19–20</td>
</tr>
<tr>
<td>Information sources</td>
<td>7</td>
<td>Describe all information sources (e.g. databases with dates of coverage, contact with study authors to identify additional studies) in the search and date last searched</td>
<td>85</td>
</tr>
<tr>
<td>Search</td>
<td>8</td>
<td>Present full electronic search strategy for at least one database, including any limits used, such that it could be repeated</td>
<td>85–7</td>
</tr>
<tr>
<td>Study selection</td>
<td>9</td>
<td>State the process for selecting studies (i.e. screening, eligibility, included in systematic review, and, if applicable, included in the meta-analysis)</td>
<td>22–3</td>
</tr>
<tr>
<td>Data collection process</td>
<td>10</td>
<td>Describe method of data extraction from reports (e.g. piloted forms, independently, in duplicate) and any processes for obtaining and confirming data from investigators</td>
<td>24</td>
</tr>
<tr>
<td>Data items</td>
<td>11</td>
<td>List and define all variables for which data were sought (e.g. PICOS, funding sources) and any assumptions and simplifications made</td>
<td>95–102</td>
</tr>
<tr>
<td>Risk of bias in individual studies</td>
<td>12</td>
<td>Describe methods used for assessing risk of bias of individual studies (including specification of whether this was done at the study or outcome level, and how this information is to be used in any data synthesis)</td>
<td>25</td>
</tr>
<tr>
<td>Summary measures</td>
<td>13</td>
<td>State the principal summary measures (e.g. risk ratio, difference in means)</td>
<td>26</td>
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<tr>
<td>Synthesis of results</td>
<td>14</td>
<td>Describe the methods of handling data and combining results of studies, if done, including measures of consistency (e.g. I²) for each meta-analysis</td>
<td>26–7</td>
</tr>
<tr>
<td>Risk of bias across studies</td>
<td>15</td>
<td>Specify any assessment of risk of bias that may affect the cumulative evidence (e.g. publication bias, selective reporting within studies)</td>
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<tr>
<td>Additional analyses</td>
<td>16</td>
<td>Describe methods of additional analyses (e.g. sensitivity or subgroup analyses, meta-regression), if done, indicating which were prespecified</td>
<td>27</td>
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<tr>
<td>Section/topic</td>
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<td>Checklist item</td>
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<tr>
<td><strong>Results</strong></td>
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<tr>
<td>Study selection</td>
<td>17</td>
<td>Give numbers of studies screened, assessed for eligibility, and included in the review, with reasons for exclusions at each stage, ideally with a flow diagram</td>
<td>29</td>
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<tr>
<td>Study characteristics</td>
<td>18</td>
<td>For each study, present characteristics for which data were extracted (e.g. study size, PCOS, follow-up period) and provide the citations</td>
<td>Summary presented given number of studies: pp. 30–2</td>
</tr>
<tr>
<td>Risk of bias within studies</td>
<td>19</td>
<td>Present data on risk of bias of each study and, if available, any outcome level assessment (see item 12)</td>
<td>Summary presented given number of studies: pp. 32–3</td>
</tr>
<tr>
<td>Results of individual studies</td>
<td>20</td>
<td>For all outcomes considered (benefits or harms), present, for each study: (a) simple summary data for each intervention group (b) effect estimates and CIs, ideally with a forest plot</td>
<td>Only done for SF-36 and MMPI analyses due to number of studies: pp. 42–66</td>
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<tr>
<td>Synthesis of results</td>
<td>21</td>
<td>Present results of each meta-analysis done, including CIs and measures of consistency</td>
<td>36–69</td>
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<tr>
<td>Risk of bias across studies</td>
<td>22</td>
<td>Present results of any assessment of risk of bias across studies (see item 15)</td>
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<tr>
<td>Additional analysis</td>
<td>23</td>
<td>Give results of additional analyses, if done [e.g. sensitivity or subgroup analyses, meta-regression (see item 16)]</td>
<td>36–69</td>
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<tr>
<td><strong>Discussion</strong></td>
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<td>Summary of evidence</td>
<td>24</td>
<td>Summarise the main findings including the strength of evidence for each main outcome; consider their relevance to key groups (e.g. health-care providers, users and policy-makers)</td>
<td>67–9</td>
</tr>
<tr>
<td>Limitations</td>
<td>25</td>
<td>Discuss limitations at study and outcome level (e.g. risk of bias) and at review level (e.g. incomplete retrieval of identified research, reporting bias)</td>
<td>68–9</td>
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<tr>
<td>Conclusions</td>
<td>26</td>
<td>Provide a general interpretation of the results in the context of other evidence, and implications for future research</td>
<td>69</td>
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<td><strong>Funding</strong></td>
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<tr>
<td>Funding</td>
<td>27</td>
<td>Describe sources of funding for the systematic review and other support (e.g. supply of data); role of funders for the systematic review</td>
<td>HTA review</td>
</tr>
</tbody>
</table>

HTA, health technology assessment; NA, not applicable.
Appendix 5

Included papers


Addington-Hall J, Walker L, Jones C, Karlsen S, McCarthy M. A randomised controlled trial of postal versus interviewer administration of a questionnaire measuring satisfaction with, and use of, services received in the year before death. *J Epidemiol Community Health* 1998;52:802–7.


Allison T, Ahmad T, Brammah T, Symmons D, Urwin M. Can findings from postal questionnaires be combined with interview results to improve the response rate among ethnic minority populations? Ethn Health 2003;8:63–9.


Cohen DB. Relation of anxiety level and defense style to frequency of dream recall estimated by different methods. Psychophysiology 1968:224.

Collins FE, Jones KV. Investigating dissociation online: validation of a web-based version of the dissociative experiences scale. J Trauma Dissociation 2004;5:133–47.


Franke GH. Effects of computer administration on the Symptom Checklist (SCL-90-R) with a special focus on the item sequence. *Diagnostica* 1999;45:147–53.


Frost NA, Sparrow JM, Hopper CD, Peters TJ. Reliability of the VCM1 Questionnaire when administered by post and by telephone. *Ophthalmic Epidemiol* 2001;8:1 November.


O’Dell WF. Personal interviews or mail panels? J Mark 1962;26:34–9.


Schulberg D. The MMPI is less sensitive to the automated testing format than it is to repeated testing: item and scale effects. *Comput Hum Behav* 1988;4:285–98.


Van Den Kerkhof EG, Parlow JL, Goldstein DH, Milne B. In Canada, anesthesiologists are less likely to respond to an electronic, compared to a paper questionnaire. *Can J Anesth* 2004;51:449–54.


Appendix 6

Papers excluded at second stage


Bratton GR, Newsted PR. Response effects and computer-administered questionnaires: the role of the entry task and previous computer experience. *Behav Inform Tech* 1995;14:300–12.


Epstein JF; Barker PR, Kroutil LA. Mode effects in self-reported mental health data. *Publ Opin Q* 2001;65:529–49.


Holbrook AL, Green MC, Kroasnick JA. Telephone versus face-to-face interviewing of national probability samples with long questionnaires: comparisons of respondent satisficing and social desirability response bias. Publ Opin Q 2003;67:79–125.


Holst E, Holstein BE. [Sociomedical survey among the elderly in 10 EEC countries. An analysis based on non-respondents to a questionnaire survey of the population 70–95 years of age living in 4 Danish communities.] Ugeskr Laeger 1990;152:225–7.


Kartashov AI, Buzin VV, Glazunov IS, Abol’ian LV. Results of testing the basic questionnaire SINDI in the evaluation of prophylaxis program development. Sov Zdravookh 1991:45–9.


Miles EW, King WC. Gender and administration mode effects when pencil-and-paper personality tests are computerised. *Educ Psychol Meas* 1998;58:68–76.


Roccaforte WH, Burke WJ, Bayer BL, Wengel SP. Validation of a telephone version of the mini-


Rose G, McCartney P, Reid DD. Self-administration of a questionnaire on chest pain and

Rosenfeld P, Doherty LM, Vicino S, Kantor J. Attitude assessment in organisations: testing three

Rosenfeld R, Dar R, Anderson D, Kobak KA. A computer-administered version of the Yale-

Ross SJ, Young CM. Mail versus Web questionnaires in municipal recreation settings: a

Rossier P, Wade DT. The Guy’s Neurological Disability Scale in patients with multiple sclerosis: a

Rudiger K. Is computer assessment of obsession and compulsion applicable in obsessive–
compulsive disorder? Preliminary results using the Hamburg Obsession Compulsion Inventory-

Sarrazin MS, Hall JA, Richards C, Carswell C. A Comparison of computer-based versus pencil-

Sasyniuk NGH, Hollinshead TM, Hollinshead RL, Mohtadi RM. A prospective pilot study
comparing paper based and computer based outcome instruments. Poster Session. *Clin J Sport

Schaik PV, Ahmed T, Suvakovic N, Hindmarsh JR. Effect of an educational multimedia prostate

Schuldburg D. Varieties of inconsistency across test occasions: effects of computerised test

Schumacher J, Hinz A, Hessel A, Brahler E. On the comparability of internet-based and paper-
and-pencil surveys: a study with the Questionnaire of Recalled Parental Rearing Behaviour.

Schwartz NS, Mebane DL, Malony HN. Effects of alternative modes of administration on

Schwenkmezger P, Hank P. Paper-and-pencil versus computer-assisted presentation of state–trait

Searles JS, Helzer JE, Rose GL, Badger GI. Concurrent and retrospective reports of alcohol
consumption across 30, 90 and 366 days: interactive voice response compared with the timeline

Sears RR. Comparison of interviews with questionnaires for measuring mothers’ attitudes

questionnaire with the IUATLD written questionnaire for measuring asthma prevalence. *Clin Exp

Shrier LA, Shih MC, Beardslee WR. Affect and sexual behaviour in adolescents: a review of
the literature and comparison of momentary sampling with diary and retrospective self-report


Appendix 7

Description of Minnesota Multiphasic Personality Inventory scales

The Minnesota Multiphasic Personality Inventory scales

Scale 1 — Hypochondriasis
This scale was originally developed to identify patients who manifested a pattern of symptoms associated with the label of hypochondria. All the items on this scale deal with subjects who are unrealistically concerned with bodily complaints. Scale 1 is designed to assess a neurotic concern over bodily functioning. A person who is actually physically ill will obtain only a moderate score on the hypochondriasis scale. These people will endorse their legitimate physical complaints, but will not endorse the entire range of vague physical complaints included in this scale.

Scale 2 — Depression
This scale focuses on lack of hope in the future, a general dissatisfaction with one's own life situation and poor morale. Low scores signify a general unhappiness with life, but high scores indicate clinical depression.

Scale 3 — Hysteria
This scale looks at hysterical reaction to stressful situations. People will often have a ‘normal’ facade and then break down when faced with high ‘trigger’ levels of stress. High scores on this scale indicate people that are more intelligent, better educated and from a higher social class. Women have predominantly scored higher than men on this scale.

Scale 4 — Psychopathic deviation
This scale measures social deviation and looks at lack of acceptance of authority and amorality. Higher scores on this scale are generally achieved by adolescents. This scale was originally developed to identify patients diagnosed as having a psychopathic personality. Scale 4 can be thought of as a measure of rebelliousness; a higher score will indicate rebellion and lower scores indicate an acceptance of authority.

Scale 5 — Masculinity–femininity
This scale was originally developed to identify homosexuality, but was unable to do so accurately. The masculinity–femininity scale is now used to measure how strongly an individual identifies with the traditional (pre-1960s) masculine or feminine role, intelligence, education and socioeconomic status.

Men on average tend to obtain higher scores on the masculinity–femininity scale. High scores are extremely uncommon among females. If a high score is achieved it can generally indicate a rejection of the traditional female role.

Scale 6 — Paranoia
This scale looks at paranoid symptoms such as suspiciousness, grandiose self-concepts, excessive sensitivity, ideas of reference, feelings of persecution and rigid opinions and attitudes. A high
score on the paranoia scale indicates that the subject has strong, irrational suspicions and overestimates his or her own self-importance.

**Scale 7 — Psychasthenia**
This scale was originally designed to look at symptoms such as compulsion, obsessions, excessive doubt and unreasonable fears. Psychasthenia indicates conditions such as obsessive-compulsive disorder. The scale also highlights difficulties in concentration, self-criticism, abnormal fears and guilty feelings. High scores on the psychasthenia scale highlight that the subject may be tense and anxious and may have obsessive thoughts or compulsive behaviours.

**Scale 8 — Schizophrenia**
This scale assesses a wide variety of content areas, including bizarre thought processes and peculiar perceptions, social alienation, poor familial relationships, difficulties in concentration and impulse control, lack of deep interests, disturbing questions of self-worth and self-identity, and sexual difficulties. High scores on this scale indicate that the subject is withdrawn, may experience distortions of reality and can tend to act bizarrely.

**Scale 9 — Hypomania**
This scale tests for elevated mood, accelerated speech and motor activity, irritability, flight of ideas and brief periods of depression. A participant who achieves a high score on is likely to be outgoing, impulsive, overly active and excited.

**Scale 0 — Social introversion**
This scale looks at a person’s inclination to withdraw from social contacts and responsibilities; thus, it will assess how shy or outgoing a person is. Hence, if a high score is achieved it indicates the subject is withdrawn, shy, inhibited and unassuming.

**Validity scales**

The authors also developed four validity scales to improve the overall accuracy of the measure, detect ‘deviant test-taking attitudes’ and gauge the accuracy of the other scales.

**The ‘cannot say’ scale – ?**
The ‘cannot say’ scale is the frequency of the number of items omitted or which have been marked both true and false on the whole outcome measure. If the scale has large number of missing items this can call into question the scores on all the other scales. The MMPI manual suggests that participants with 30 or more omitted items should be considered invalid and not interpreted. High scores on this scale can also indicate that the subject is indecisive.

**The L scale**
Originally called the ‘lie’ scale, this attempted to assess naive or unsophisticated attempts by people to present themselves in an overly favourable light. In terms of scoring, people who obtain high L scores are not willing to admit even minor shortcomings, hence, are deliberately trying to present themselves in a more positive way. People who are better educated and more sophisticated people from a high social class tend to score lower on the L scale.

**The F scale**
This is the deviant or rare response scale. The scale will analyse the items which are rarely endorsed by normal people. If less than 10% of the normal population sanction the item, but you endorse it, your F score would increase. For instance ‘all laws should be eliminated.’
The F scale has three vital functions:

1. It is an index of test-taking attitude and is useful in detecting deviant response sets (i.e. faking good or faking bad).
2. If one can rule out profile invalidity, the F scale is a good indicator of degree of psychopathology, with higher scores suggesting greater psychopathology.
3. Scores on the F scale can be used to generate inferences about other characteristics and behaviours.

The K scale

The K scale was designed to analyse more subtle distortion of response, particularly clinically defensive response. The K scale was constructed by comparing the responses of a group of people who were known to be clinically deviant but who produced normal MMPI profiles with a group of normal people who produced normal MMPI profiles (no evidence of psychopathology in both). The K scale was subsequently used to alter scores on other MMPI scales. It was reasoned that people with high K values give scores on other scales which are too low, for instance if the participant achieves a high K score it will indicate that the subject is defensive and attempting to obscure symptoms. K is used to adjust the scores on other scales.
Appendix 8

Description of Short Form questionnaire-36 items health scales

**Physical functioning**

Measures how able a responder is to perform physical tasks without limitations due to health.

**Role physical**

Measures due to physical health, a responder has problems with work or other daily activities.

**Bodily pain**

Measures the severity and level of limitation due to bodily pain.

**General health perception**

Measures overall health.

**Vitality**

Measures energy levels and fatigue.

**Social functioning**

Measures the level of interference with social activities due to physical or emotional problems.

**Role emotional**

Measures how much emotional problems impact on work or daily activities.

**Mental health**

Measures levels of individual mental health.
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Professor of Community Psychiatry, Centre for Mental Health, Imperial College London

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<th>Affiliation</th>
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<tr>
<td>Chair</td>
<td>Professor Martin Underwood, Professor of Primary Care Research, Warwick Medical School, University of Warwick</td>
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<tr>
<td>Chair</td>
<td>Professor Caroline Watkins, Professor of Stroke and Older People’s Care, Chair of UK Forum for Stroke Training, Stroke Practice Research Unit, University of Central Lancashire</td>
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<tr>
<td>Deputy Chair</td>
<td>Dr Duncan Young, Senior Clinical Lecturer and Consultant, Nuffield Department of Anaesthetics, University of Oxford</td>
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<td>Observers</td>
<td>Dr Tom Foulks, Medical Research Council</td>
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<tr>
<td>Observers</td>
<td>Dr Kay Pattison, Senior NIHR Programme Manager, Department of Health</td>
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**HTA Clinical Evaluation and Trials Board**

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<tr>
<td>Chair</td>
<td>Professor Sallie Lamb, Director, Warwick Clinical Trials Unit, Warwick Medical School, University of Warwick and Professor of Rehabilitation, Nuffield Department of Orthopaedic, Rheumatology and Musculoskeletal Sciences, University of Oxford</td>
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<tr>
<td>Deputy Chair</td>
<td>Professor Jenny Hewison, Professor of the Psychology of Health Care, Leeds Institute of Health Sciences, University of Leeds</td>
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<td>Programme Director</td>
<td>Professor Tom Walley, CBE, Director, NIHR HTA programme, Professor of Clinical Pharmacology, University of Liverpool</td>
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**Members**

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<td>Chair</td>
<td>Professor Keith Abrams, Professor of Medical Statistics, Department of Health Sciences, University of Leicester</td>
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<td>Chair</td>
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<td>Chair</td>
<td>Professor Julia M Brown, Director, Clinical Trials Research Unit, University of Leeds</td>
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<tr>
<td>Chair</td>
<td>Professor Alistair Burns, Professor of Old Age Psychiatry, Psychiatry Research Group, School of Community-Based Medicine, The University of Manchester &amp; National Clinical Director for Dementia, Department of Health</td>
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<tr>
<td>Chair</td>
<td>Professor Jennifer Burr, Director, Centre for Healthcare Randomised trials (CHART), University of Aberdeen</td>
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<tr>
<td>Chair</td>
<td>Professor Linda Davies, Professor of Health Economics, Health Sciences Research Group, University of Manchester</td>
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<tr>
<td>Chair</td>
<td>Professor Simon Gilbody, Prof of Psych Medicine and Health Services Research, Department of Health Sciences, University of York</td>
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<td>Professor Dyfrig Hughes, Professor of Pharmacoeconomics, Centre for Economics and Policy in Health, Institute of Medical and Social Care Research, Bangor University</td>
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<tr>
<td>Chair</td>
<td>Professor Paul Jones, Professor of Respiratory Medicine, Department of Cardiac and Vascular Science, St George’s Hospital Medical School, University of London</td>
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<td>Chair</td>
<td>Professor Khalid Khan, Professor of Women’s Health and Clinical Epidemiology, Barts and the London School of Medicine, Queen Mary, University of London</td>
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<tr>
<td>Chair</td>
<td>Professor Helen Rodgers, Professor of Stroke Care, Institute for Ageing and Health, Newcastle University</td>
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<tr>
<td>Chair</td>
<td>Professor Ken Stein, Professor of Public Health, Peninsula Technology Assessment Group, Peninsula College of Medicine and Dentistry, Universities of Exeter and Plymouth</td>
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<tr>
<td>Observers</td>
<td>Ms Kate Law, Director of Clinical Trials, Cancer Research UK</td>
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<tr>
<td>Observers</td>
<td>Dr Morven Roberts, Clinical Trials Manager, Health Services and Public Health Services Board, Medical Research Council</td>
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