

Vocal Characteristics, Speech, and Behavior of Telephone Interviewers

by

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Abstract

Growing rates of nonresponse to telephone surveys have the potential to contribute to nonresponse error, and interviewers contribute differentially to nonresponse. Why do some telephone interviewers have better response rates than others? What should interviewers be trained to say or do to improve their performance in recruitment—and how do these recommendations differ from current practice?

This dissertation uses three studies to answer these questions. A “Practitioners’ Survey” among individuals responsible for hiring and training telephone interviewers found that practitioners place tremendous importance on the first impression an interviewer gives to sample members, including whether s/he sounds confident, competent, and professional, but far less importance on an interviewer sounding natural or unscripted.

This widespread belief in the industry contrasts sharply with results from the “Listeners’ Study,” which exposed over 3,000 web survey respondents to brief excerpts of interviewer speech from audio recorded survey introductions and asked them to rate twelve personality characteristics of the interviewer. First impressions of confidence, competence, professionalism and other traits had no association with the actual outcome of the call, while ratings of “scriptedness” were significantly negatively associated with the likelihood of agreement. At the same time, ratings of positive personality traits were positively associated with

predictions by different groups of raters as to the outcome of the call. Further, significant relationships were found between measured speech rate and fundamental frequency in the excerpts and characteristic ratings.

Beyond first impressions, the “Tailoring Study” uncovered a critical and trainable behavior of successful telephone interviewers over the course of introductions. Using detailed coding of 626 introduction transcripts, interviewers’ responsiveness to specific concerns such as “I don’t have time” or “What is this about?” and conversation starters by potential respondents or telephone “answerers” is analyzed across contacts with three outcomes: agree, refusal, and scheduled callback. Results show that interviewers are most responsive to answerers in calls that result in a scheduled callback and least responsive in refusals.

Practical applications for telephone interviewer training are discussed, including suggested approaches to both “red flags” indicating an imminent hang-up and “green lights” suggesting likely agreement.

Chapter 1: A Proposed Conceptual Model for the Association Between Interviewers' Speech and Vocal Characteristics and Success

1.1 Problem Statement, Significance, and Research Aims

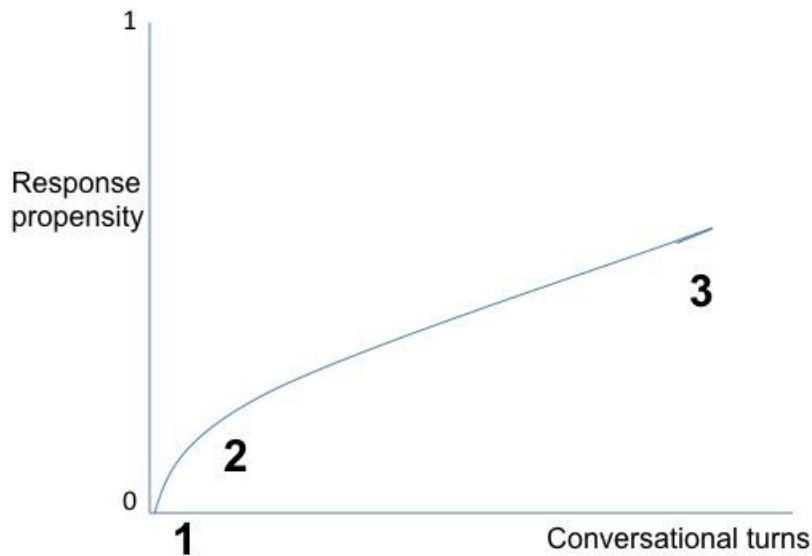
Nonresponse to telephone surveys has the potential to bias survey estimates (Groves, Presser, and Dipko 2004), which in turn can have policy and programmatic implications. It has been demonstrated (Oksenberg and Cannell 1988) that some telephone interviewers have higher response rates—that is, more success recruiting sample members to participate—than others. Identifying vocal characteristics and techniques of successful telephone interviewers promises to have a potentially broad impact on data quality, by allowing for more targeted screening and training of interviewers with the aim of reducing nonresponse, which can in turn benefit any disciplines in which telephone survey data is used.

Literature from both survey methodology (Oksenberg, Coleman, and Cannell 1986) and telemarketing (Ketrow 1990) has found that a pleasing or attractive voice in the initial seconds of a phone call is imperative in extending the interaction. Further, Ketrow (1990) discusses the importance of giving an initial impression of competence, and Lois Oksenberg and colleagues (Oksenberg and Cannell 1988; Oksenberg, Coleman, and Cannell 1986;) find that judges' ratings of competence were associated with phone interviewers' success. This is not to imply that, in the context of a survey interview introduction, having a pleasing,

competent-sounding voice in the opening statement is enough to guarantee success. However, a voice which gives listeners a positive first impression may provide an interviewer with an entrée into a longer conversation and, by extension, more opportunities to tailor their introduction to a specific sample member. The theory of tailoring (Groves and Couper 1998) states that successful interviewers tend to adapt their recruitment approach to particular sample members, for example by modifying their attire, using arguments that they believe will resonate with an individual, or changing their style of speech. Most studies of tailoring have focused on face-to-face studies, although Maynard, Freese, and Schaefer (2010) find that telephone interviewers who fail to adapt their approach (for example, by being “presumptive” when a “cautious” approach is warranted) have less success. Introductions in telephone studies are admittedly brief, and interviewers are limited in this setting not only in the information they can glean from the sample members who answer the phone (referred to hereafter as phone “answerers”), but also in the ways they are able to adapt their approach. Displaying responsiveness to sample members’ concerns is one method of adapting that may occur in a telephone survey context.

This dissertation proposes a three-stage model of successful recruiting by a telephone interviewer, illustrated in Figure 1.1. The model is based on an interdisciplinary framework that draws on theories of survey methodology (tailoring), sociolinguistics (including theories of politeness and conversational grounding), and psychology (the theory of person perception). Some phone

answerers will be determined not to participate, and no aspect of the interviewer's voice or speech can persuade them otherwise. Notwithstanding the proclivities of these individuals, I posit that the initial impression of an interviewer as warm and competent can allow him or her to get a "foot in the door," so to speak. Groves and Couper (1998) suggest that interviewers who are able to prolong interactions will be able to obtain more information for effective tailoring. Beyond the initial impression, displays of responsiveness and competence can increase the likelihood that a sample member will participate.



- 1: Some answerers will not respond— no matter what an interviewer says or how s/he sounds.
- 2: Initial presentation of a warm, competent voice can get an interviewer "over the hump" of the interaction.
- 3: Responsiveness for the remainder of the introduction will improve response likelihood.

Figure 1.1: A Proposed Conceptual Model of Interviewer Behavior

This chapter reviews the existing literatures on person perception and tailoring. While both of these ideas are relevant to survey nonresponse, they have not been examined in combination; I propose that they are, in fact,

interrelated and should all be considered in discussions of telephone survey response. I will integrate these concepts to develop a model from the perspective of the interviewer (as opposed to the answerer) that can begin to explain differences in success between interviewers based on their vocal characteristics, speech, and behaviors.

1.2 Overview of Studies

This dissertation begins with results from a study I conducted measuring the impact of initial judgments of telephone interviewers' personality and vocal characteristics on their success at turning answerers into respondents. In particular, the study asked 3,403 judges to rate speech and personality traits of interviewers based on audio recordings presented online; this was administered as a Web survey in which the judges were respondents. These ratings were used to predict the outcome of the contact. My second study assessed how interviewers' subsequent responsiveness to sample members' concerns and conversation starters *after* the initial impression affect their success. This study relied on coded interactions to assess the interviewers' responsiveness. The third study collected survey practitioners' beliefs about the importance of interviewers' speech, voices, and behaviors in obtaining interviews. The findings provide context and a point of contrast for results from the other two studies.

Specifically, the first study collected ratings of interviewers on the following traits: enthusiastic, happy, genuine, pleasant to listen to, confident, professional, competent, knowledgeable, natural-sounding, irritating, uncertain, and scripted. This study tested the hypothesis that ratings of the first nine of these

characteristics, either alone or in combination, predicted interviewer success as measured by the actual outcome of the recorded contact. A secondary hypothesis tested is that these traits will cluster into two factors representing warmth and competence—two dimensions widely assumed to be crucial to person perception (Asch 1946). Judges were also asked to predict the likelihood that a contact resulted in agreement to test the hypotheses that these predictions are associated with ratings of interviewer characteristics, as well as with the actual outcome of a contact. A further question explored in the first study is the association of vocal characteristics to these ratings. I hypothesize that vocal characteristics associated with competence, such as a faster rate of speech and lower fundamental frequency (Brennan and Williams 1995; Ketrow 1990), will correlate with judges' ratings of competence, confidence, knowledge and professionalism. Additionally, vocal characteristics associated with warmth, such as a higher fundamental frequency, will correlate with judges' ratings of enthusiasm, happiness, how genuine the voice sounds, and how pleasant the voice is to listen to.

The second study explored which interviewer behaviors and vocal characteristics predict a contact outcome of agreement. I hypothesize that interviewers who provide relevant responses to answerer concerns or remarks, and/or adapt their speech to the individual situation (for example, by slowing down their speech for an answerer who is struggling to understand), will be more successful than interviewers who do not.

1.3 Literature Review

Nonresponse to telephone surveys has been increasing steadily over the past 25 years (Curtin, Presser, and Singer 2005). Declining response rates have the potential to increase nonresponse error, as a greater number of nonrespondents can increase the opportunity for differences to be observed between respondents and nonrespondents (Groves, Presser, and Dipko 2004; Teitler, Reichman, and Sprachman 2003). Further, nonresponse rates vary by interviewer (Morton-Williams 1993; Oksenberg and Cannell 1988; O’Muircheartaigh and Campanelli 1999; Snijkers, Hox, and de Leeuw 1999). Oksenberg and Cannell (1988) find that even when telephone interviewers read identical scripts, their refusal rates ranged from 6 percent to 42 percent. Attitudinal and behavioral differences between face-to-face interviewers are found in studies by Campanelli, Sturgis, and Purdon (1997); Durrant et al. (2010); Hox and de Leeuw (2002); Pondman (1998); and Snijkers, Hox, and de Leeuw (1999). Blom, de Leeuw, and Hox (2010) find that variation in response rates between countries can be partially attributed to differences in interviewer behavior. Uncovering the characteristics and behaviors of successful interviewers can help to reduce nonresponse.

1.3.1 *Vocal Characteristics and Survey Response*

In contrast to face-to-face interviewers, telephone survey interviewers have only two tools at their disposal in the effort to persuade answerers to participate: what they say (speech) and how they say it (vocal characteristics). Vocal characteristics such as rate and pitch are considered more important to persuasion in situations where a listener has less rather than more involvement

and investment in the outcome (Gelinus-Chebat and Chebat 1992; Smith and Shaffer 1995). This distinction may be applicable to the survey invitation, a situation in which potential respondents tend to have low involvement.

A small body of literature (e.g., Groves, O'Hare, Gould-Smith, Benkí and Maher 2007; Oksenberg and Canell 1988; Oksenberg, Coleman, and Cannell 1986; Sharf and Lehman 1984) finds relationships between vocal characteristics of interviewers in telephone survey introductions and interviewer success. As indicated in Table 1.1 below, the methodologies differed across these studies.

Table 1.1: Summary of Studies on Interviewer Vocal Characteristics

Study	Methodology
Sharf and Lehman (1984)	Eighteen students rated vocal and personal characteristics of interviewers based on recordings of the first 30 seconds of one call for each of six interviewers with varying historical response rates. Acoustic measures were also assessed.
Oksenberg, Coleman, and Cannell (1986)	Two groups of raters (18 students and 40 volunteers) rated vocal and personal characteristics of interviewers based on recordings of the first 30 seconds of one call for each of six interviewers with varying historical response rates.
Oksenberg and Cannell (1988)	Four judges rated vocal and personal characteristics for 55 interviewers with varying historical response rates; acoustic measures were also assessed.
Huefken and Schaefer (2003)	Judges rated speech and vocal characteristics from 219 interview introductions by 51 student interviewers with varying success rates.
van der Vaart., Ongena, Hoogendoorn, and Dijkstra (2005)	Twelve judges rated vocal and personal characteristics, as well as their own willingness to participate, for two recordings by each of 31 student interviewers. Cooperation rates for "interviewers" were based on 2,155 answerers. Acoustic measures were also assessed.
Groves and Benkí (2006)	Eight raters listened to the interviewer's first conversational turn for 300 contacts by 22 interviewers and rated speech, vocal, and personal characteristics, as well as the likelihood that "someone" would agree to the request.
Groves et al. (2007)	Eight judges rated vocal and personal characteristics of 58 interviewers, as well as the likelihood that "someone" would agree to the request. Acoustic measures were also assessed.
Steinkopf, Bauer, and Best (2010)	One hundred fifty-nine judges rated personal characteristics of 56 student interviewers, as well as their own willingness to participate. Acoustic measures were also assessed.

The methods summarized above point to challenges in conducting research in this area. Often the independent variables are judges' ratings of an interviewer's pitch, speech rate, or other vocal characteristics. When such ratings are used, small sample sizes tend to be the norm; for example, the two earliest studies were each based on six recordings. Studies with larger sample sizes, such as Huefken and Schaefer (2003), Steinkopf, Bauer, and Best (2010), and van der Vaart et al. (2005), are based on the work of student (rather than professional) interviewers, limiting the applicability of findings. In addition, while subjective ratings of vocal characteristics such as pause frequency and fluency demonstrate listeners' *impressions* of interviewers, none of the studies mentioned captured these variables objectively (for example, by computing rates of fillers such as "um" and "uh"). Further, the content of interviewers' speech is not considered in any existing work on interviewer vocal characteristics. Certainly, while capturing the content of an introduction is labor-intensive, including it in analyses will give a more complete picture of the introduction. Existing studies do not take conversational interaction into account; judges' ratings are based on interviewers' speech and vocal characteristics, to the exclusion of utterances by potential respondents. Finally, dependent variables assessed in existing studies are either interviewers' historical response rates, judges' own willingness to comply, or judges' beliefs that someone will comply; no study has yet associated vocal characteristics with actual contact outcomes.

Despite these limitations and variations in measures and sample sizes, some vocal characteristics have consistently correlated with interviewer success. In

general, successful interviewers have been ones who spoke louder (Oksenberg and Cannell 1988; Oksenberg, Coleman, and Cannell 1986, van der Vaart et al. 2005), faster (Oksenberg and Cannell 1988; Oksenberg, Coleman, and Cannell 1986), and with more falling intonation (Oksenberg and Cannell 1988; (Sharf and Lehman 1984). In addition, success has been shown to be correlated with both higher mean fundamental frequency (Sharf and Lehman 1984) and higher perceived pitch (Oksenberg, Coleman, and Cannell 1986), as well as variable fundamental frequency (Groves et al. 2007; Sharf and Lehman 1984) and variable pitch (Oksenberg, Coleman, and Cannell 1986). (The terms “pitch” and “fundamental frequency” are often used interchangeably, but a necessary distinction is that fundamental frequency is an acoustic measure of vocal chord vibrations, while pitch is a listener’s perception of frequency, or how “high” or “low” a voice sounds.) More recent studies find nonlinear relationships between success and speed (Groves et al. 2007; Steinkopf, Bauer, and Best 2010).

1.3.2 Formation of First Impressions

One critical question concerns what underlies these associations; what is it about an interviewer who speaks faster or with more variable pitch that leads to success, especially given the limited amount of exposure an answerer has to the interviewer’s voice? Two dimensions of person perception, “warmth” and “competence,” have been shown to be universally relevant to the development of impressions of other people (Asch 1946; Fiske, Cuddy, and Glick 2007; Kelley 1950; Rosenberg, Nelson, and Vivekanathan 1968). Fiske, Cuddy, and Glick (2007) explain these perceptions from an evolutionary perspective: warmth, or

determining whether a stranger is a potential friend or foe, is by necessity the first judgment made of another person, and it is made, to some degree, in under a second. Judgments about competence are secondary to warmth judgments; from an evolutionary perspective, these judgments involved determining whether an encountered stranger had the ability to act on any threats or helping tendencies that may have been suggested by the warmth judgment (Fiske, Cuddy, and Glick 2007). Although these evolutionary needs for quick judgments of others have become largely obsolete, the importance of first impressions on subsequent interactions, behavior, and expectations persists in many contexts (Jones 1990).

Person perception is generally assumed to occur in two stages: first, an automatic “gut feeling,” and then, a more controlled process (Ambady, Krabbenhoft, and Hogan 2006). First impressions can be developed very quickly. Ambady and colleagues (Ambady, Krabbenhoft, and Hogan 2006; Ambady and Rosenthal 1992) collect judgments based on what they refer to as “thin slice” exposures of as short as 20 seconds to a voice and found that “snap” judgments, particularly on an interpersonal dimension, are the same as those made based on more exposure.

Still briefer exposures to voices are used by Hecht and LaFrance (1995) in their study of telephone operators. In this study, exposures of only five seconds allowed judges to make ratings that correlated with operators’ job performance. These authors found that ratings of operators as enthusiastic, sympathetic, confident, professional, and friendly were highly correlated, and concluded that judges used some “underlying dimension of positivity or vocal attractiveness to

make their evaluations” (p. 2095). Operators who were ranked highly on these characteristics tended to also rate highly on the job performance measure at issue—speed—and were also described as having less monotonous voices, and clearer pronunciation. Despite the difference in context between calls to phone operators and calls from survey interviewers, the finding that “a positive tone in the initial seconds may set the stage for a smooth, coordinated interchange of information” (p. 2095) is relevant to the introduction to phone surveys: a voice that gives an immediately positive impression may have the potential to make even a request for participation sound appealing.

There is support for the idea that positive first impressions are critical in a telephone, and specifically a telephone survey, context. Oksenberg, Coleman, and Cannell (1986) emphasize the importance of an initially appealing voice in a survey interview introduction context, stating that “if vocal characteristics lead the respondent to perceive the interviewer as unappealing, cooperation will be less likely” (p. 99). The telemarketing literature also stresses the importance of an appealing voice in the first 10 to 15 seconds of a sales call (Ketrow 1990), and Sanchez (1985) notes that content is irrelevant if a salesperson’s voice is deemed unappealing in the initial seconds of a call. The importance of pleasing voices in brief telephone encounters is demonstrated in Bartsch’s (2009) study of voice-to-voice customer service encounters. This study finds that not only were ratings of vocally attractive service representatives’ efforts and abilities rated higher than their vocally unattractive counterparts, but *expectations* of these

attributes were rated higher based on brief initial exposures to a vocally attractive service representative.

While first impressions are formed quickly, their impact is long lasting. Judges tend to form stable impressions of a person after an initial exposure—what Zuckerman et al. (1979) call a “demeanor effect.” Zuckerman and Driver (1989) extend this effect to vocal attractiveness, finding that differences in vocal attractiveness led to different impressions of personality, but that these differences held regardless of the message presented by the speaker.

1.3.3 Voice and Personality Characteristics

Several studies in the literature on interviewer vocal characteristics suggest that ratings of personal characteristics on the two dimensions of person perception (warmth and competence) are associated with both interviewer success and vocal characteristics. These studies involve collecting ratings of several interviewer personality characteristics, which were then successfully reduced to two dimensions interpretable as “warmth” and “competence.” Characteristics on the “warmth” dimension included cheerful, friendly, enthusiastic, interested in the task, and pleasant to listen to. Characteristics assessing the “competence” dimension included educated, intelligent, and professional. Oksenberg, Coleman, and Cannell (1986) find correlations between high ratings on both of these dimensions and vocal characteristics including variation in pitch, higher pitch, loudness, a faster rate of speech, and more distinct pronunciation. Van der Vaart et al. (2005) find that interviewers rated highly on characteristics on the warmth dimension (enthusiastic, friendly, social,

personal, not aloof, and polite) tended to have higher and more variable pitch, while those rated highly on “competence” characteristics such as overriding, self-assured, objective, and professional tended to have lower pitch.

Both Oksenberg, Coleman, and Cannell (1986) and van der Vaart et al. (2005) find that high ratings on a “warmth” dimension correlated with ratings of judges’ willingness to participate. This aligns with Morton-Williams’ (1993) finding that warm or “likable” interviewers increased perceived benefits to potential respondents and improved participation rates, and also with Cialdini’s (1984) “Liking” Principle of Compliance: people are more likely to comply with a request from someone they like.

Oksenberg and Cannell (1988) find that ratings of how pleasant an interviewer was to listen to were positively related to several subjective ratings of vocal characteristics, including “stress or emphasis of words and phrases,” “flow of words,” and “spontaneity.” Interviewers who were “pleasant to listen to” had greater success, indicating that an underlying dimension of vocal attractiveness is related to survey response. Judgments of vocal attractiveness have been shown in a number of studies (Bartsch 2009; Berry, Hansen, Landry-Pester, and Meier 1994; Zuckerman and Miyake 1993) to be highly reliable between raters and also to be positively correlated with judgments of desirable personality traits on the dimensions of warmth and competence, including warmth, honesty, competence, leadership, dominance (Berry et al. 1994), achievement, and likability (Zuckerman and Driver 1989). Zuckerman and colleagues (Zuckerman and Driver 1989; Zuckerman, Hodgins, and Miyake 1990) attribute the

assignment of desirable personality traits to vocally attractive people to what they term the “what sounds beautiful is good” stereotype. The “what’s beautiful is good” stereotype, whereby physically attractive people are associated with desirable personality traits, has roots in the physical attractiveness realm; it was first demonstrated by Dion, Berscheid, and Walster (1972).

While an appealing voice contributes to positive first impressions, Zuckerman and Miyake (1993) caution that vocal attractiveness is low in salience and listeners are often not aware of it. In keeping with the dual-component model of person perception, it makes sense to also account for the role of competence in first impressions. Maddux and Rogers (1980) determine that physical attractiveness has *no* main effect on persuasion. Instead, attractive people may need expertise or supporting arguments in order to persuade others. It is not terribly far-fetched to think that the same may be true of vocal attractiveness or appeal: in and of itself it is not particularly persuasive, and competence and credibility of a speaker are also necessary for persuasion. Cialdini (1984) suggests a compliance heuristic based on the principle of authority; an authoritative requester is more likely than one without authority to have a request granted. Findings from qualitative interviews conducted as part of my preliminary studies (discussed in detail in Appendix 1) suggest that respondents cannot reliably rate broad, global characteristics of interviewers, such as “authority” or “likability.” However, respondents often use evidence of specific characteristics such as competence, confidence, enthusiasm, and genuineness to make assessments of wide-reaching characteristics such as “authority” and “likability.”

Impressions of competence and confidence, in turn, have been shown to be associated with interviewer success (Oksenberg and Cannell 1988; Oksenberg, Coleman, and Cannell 1986; Steinkopf, Bauer, and Best 2010).

1.3.4 Tailoring and Responsiveness

The initial impression of a phone interviewer's voice as warm and competent (stage 1 in the model depicted in Figure 1.1) may offer the interviewer the proverbial "foot in the door," giving the interviewer an opportunity to tailor their introduction to be more relevant to the individual by keeping a potential respondent on the phone longer. Groves and Couper (1996) name "prolonging interaction" as a key strategy of successful interviewers for this reason.

Interviewer responsiveness has been shown to be effective in persuasion. Campanelli, Sturgis and Purdon (1997) find that relevant interviewer responses to specific householder concerns, indicating adaptation, are a successful interviewer technique. Groves and McGonagle (2001) are able to quantify this association with their finding that interviewers' cooperation rates in telephone surveys improved after interviewers went through a training program to increase their use of tailoring techniques, specifically by focusing on giving relevant responses to concerns expressed by sample members. Similarly, Pondman (1998) finds a 49% refusal rate among interviewers who were trained in a responsive strategy (compared to a 60% refusal rate among interviewers who did not participate in the training): in response to refusals, rather than asking "why" or repeating the refusal ("You don't want to participate?"), interviewers were instructed to offer to call back if answerers indicated not having time at the

moment to take part in the interview, and also to “apply the provision of relevant information about features of the interview in reaction to refusals based on reasons other than time” (p. 75).

Viewed through the lens of vocal characteristics, responsiveness may be indicated by linguistic convergence, or the “mirroring” by one conversational partner of the other’s speech. According to Communication Accommodation Theory (Giles, Coupland, and Coupland 1991), convergence of communication elements is a method of reducing social distance that can have positive outcomes for conversational interactions; that is, people who mirror their conversational partner are perceived more positively by that partner than those who do not. Convergence has been demonstrated with linguistic and paralinguistic elements including speech rate, accent, and response latency (Giles, Coupland, and Coupland 1991). Chartrand and Bargh (1999) discuss this “chameleon effect,” and find that individuals were more favorably inclined toward a conversational partner who mirrored their vocal and nonvocal characteristics.

While a small body of literature has shown that telephone interviewers’ personality traits (such as being warm and competent), vocal characteristics (including a faster rate of speech and a higher fundamental frequency), and behaviors (specifically, responsiveness to answerers’ concerns) are, in fact, associated with their success, many of these studies have been fairly ad hoc, and lacking in strong theoretical grounding. This dissertation helps fill this gap in the literature by developing a conceptual model for interviewer behavior which is based on interdisciplinary ideas about communication.

Chapter 2: Examining the Role of First Impressions

2.1 Introduction

This chapter asks whether first impressions, formed in the initial seconds of a telephone interviewer's introduction, are critical in determining the outcome of the contact. I intend to answer several questions concerning first impressions of telephone interviewers:

- Which first impressions of interviewer characteristics related to “warmth” and “competence” are predictive of a successful outcome?
- How do vocal characteristics relate to first impressions?
- How accurately can listeners predict contact success based only on the initial seconds of an interviewer's introduction?
- How do listeners' first impressions of an interviewer relate to other listeners' predictions of a contact's outcome?
- How do actual relationships between first impressions and success compare or contrast with survey practitioners' ideas about what makes a successful interviewer?

While this work is largely exploratory in nature, it is structured by the following hypotheses:

H1: Ratings of nine interviewer characteristics in will cluster into two factors representing the two dimensions of person perception (warmth and competence).

H2: Judges' ratings of the likelihood that a contact resulted in agreement will be associated with the actual outcome of the contact.

H3a: Higher ratings of the nine positive characteristics will be associated with higher ratings by other judges of the likelihood that the contact resulted in agreement.

H3b: Lower ratings of the three negative characteristics will be associated with higher ratings by other judges of the likelihood that the contact resulted in agreement.

H4: Ratings of interviewer characteristics based on first impressions will be associated with greater success of a survey invitation.

H4a. Higher ratings of nine positive interviewer characteristics (enthusiastic, friendly, genuine, pleasant to listen to, confident, professional, competent, knowledgeable, and natural-sounding), either alone or in combination, will be associated with greater success of a survey invitation.

H4b. Lower ratings of three negative characteristics (scripted, uncertain, and irritating) will be associated with greater success of a survey invitation.

H5: Survey practitioners will rank telephone interviewers' personality characteristics as more important than their vocal characteristics in contributing to an interviewer's success.

H6: Practitioners will view as important those characteristics that are associated with greater actual success.

H7: Vocal characteristics associated with competence, such as a moderately fast rate of speech and lower fundamental frequency, will be associated with higher ratings in the categories of competent, confident, knowledgeable, and professional.

H8: Vocal characteristics associated with warmth, such as a higher fundamental frequency, will be associated with judges' ratings in the categories of enthusiastic, friendly, genuine, and pleasant to listen to.

This chapter begins with a description of data collected in two Web surveys. The first, referred to as the “listeners’ study,” was conducted among Internet survey panel members who answered questions after listening to brief excerpts of interviewer speech from telephone survey introductions. The second was conducted among survey practitioners and is referred to as the “practitioners’ study.” Results from the listeners’ study testing the hypotheses above are described and contrasted with results from the practitioners’ study. I finish with conclusions, a discussion of limitations, and suggestions for future research in this area.

2.2 Data and Methods

The data described in this chapter are drawn from two Web surveys. The first was a survey among 3,403 adult, English-speaking members of an Internet survey panel. The second was a smaller survey of 44 survey practitioners who are responsible for the hiring and training of survey interviewers. Respondents to the practitioners’ study represented academic, government, and for-profit survey organizations.

2.2.1 Listeners' Study: Questionnaire Development

The study referred to as the “listeners’ study” used online presentation of audio recordings, fielded by Lightspeed Research as a Web survey and designed to elicit listeners’ judgments about telephone interviewers’ personality characteristics and the success of telephone survey invitations. In this survey, 3,403 members of an online survey panel listened to interviewer speech from selected phone survey introductions (the selection process is described in Section 2.2.2). The stimuli to which listeners were exposed consisted of brief introductory statements by the interviewer, such as: “Hello, my name is ___ and I’m calling from the University of Michigan about our survey on____.” No answerer speech was included in the excerpts. (As discussed in Section 1.1, the term telephone “answerer” refers to a sample member.)

All listeners heard excerpts from ten telephone survey introductions randomly selected from a corpus containing 283 introductions by 49 different interviewers, meaning that it was possible for some listeners to hear multiple introductions by one interviewer, and for others to hear ten different interviewers. The particular group of excerpts presented to each listener contained two excerpts in each of five length categories, outlined below in table 2.2. While the same group of ten contacts could be heard by multiple listeners, assignment and order of excerpts was random so as to avoid context effects from presenting excerpts in set groups or a set order.

For the first five excerpts, listeners were asked only to predict the likelihood that the sample member would agree to participate. These are referred to as “likelihood ratings” in what follows.

For five other contacts, listeners were asked to rate the interviewer on the twelve characteristics outlined in Table 2.1 using a scale from 1 (not at all) to 6 (extremely). These are referred to as “characteristic ratings” below. Many of the characteristics rated were mentioned by respondents in the preliminary qualitative study, mentioned in Chapter One, which explored perceptions of interviewers’ voices; others have been shown in the literature (Oksenberg and Cannell 1988; van der Vaart et al. 2005) to be related to interviewer success.

Table 2.1: Rated Interviewer Characteristics

Characteristic	Rationale for including
Pleasant to listen to	Oksenberg and Cannell (1988) find an association with success.
Enthusiastic	Explored by Groves et al. (2007); find no relationship with success. Mentioned repeatedly by respondents to preliminary study described in Appendix 1.
Friendly	Rated in studies by Oksenberg and Cannell (1988) and van der Vaart et al. (2005); results mixed as to an association with success.
Genuine	Used repeatedly by respondents in the preliminary study.
Confident	Oksenberg and Cannell (1988) find an association with success. Van der Vaart et al. (2005) find “self-assured” loaded highly on an “authority” factor, which is associated with interviewers’ historical response rates.
Professional	Van der Vaart et al. (2005) find “professional” loaded highly on an “authority” factor.
Competent	Oksenberg and Cannell (1988) find a positive association between competency ratings and success.
Scripted	Groves et al. (2007) find a negative association between scriptedness and success.
Knowledgeable	Often used in the preliminary qualitative study when respondents discuss whether the interviewer is “good at her job,” a measure used by Oksenberg, Coleman, and Cannell (1986).

Characteristic	Rationale for including
Natural-sounding	All included as a check on data quality, with the assumption that judges who give high ratings for “scripted,” “confident,” and “pleasant to listen to” would choose low ratings for “natural sounding,” “uncertain,” and “irritating.”
Uncertain	
Irritating	

The full listeners’ questionnaire can be found in Appendix 2.

2.2.2 Selection of Contacts

Contacts were selected from 1,380 audio-recorded telephone survey introductions. These introductions were transcribed, coded, and acoustically measured as part of a project, funded by the National Science Foundation, which explored the impact of interviewer vocal and speech characteristics on nonresponse (Benkí, Broome, Conrad, Groves and Kreuter 2011; Conrad, Broome, Benkí, Groves, Kreuter and Vannette 2010). Contacts used in this project were selected from five telephone surveys conducted at the University of Michigan, using a sampling structure involving stratification both by interviewer and outcome. This sampling structure, described in detail in Appendix 3, ensured that the dataset contained a number of successful contacts sufficient for analysis.

All contacts were classified into five outcomes: “agree,” where the answerer agrees to participate; “refuse,” where there is an explicit refusal (for example, “I will not take the survey. Please do not call again”); “scheduled callback,” where the interviewer either schedules a time to call back or asserts that she will call again; “hang up,” where the answerer hangs up but never clearly refuses; and “other.”

A subset of the 1,380 contacts described above was used for the listeners’ study described here. To facilitate comparisons (particularly in analyses of vocal

characteristics such as pitch), only introductions by female interviewers were selected. In addition, contacts in the original dataset from the Gujarati Community Study, a study conducted among members of a distinct linguistic and cultural community, were excluded from selection because they often included labored communication between interviewers and answerers. Finally, contacts where the answerer hangs up during or directly following the interviewer's first speaking turn were excluded, using the rationale that these are "hard-core nonrespondents" who are determined not to become respondents, and nothing in the interviewer's voice or speech can convince them otherwise.

After these criteria were applied, 674 invitations remained. Only agrees and refusals were included in this analysis, so as to foster a cleaner comparison. Three hundred twenty-four contacts with outcomes that were not clearly "agree" or "refusal" were excluded (276 had the outcome "scheduled callback," nine had the outcome "hang-up," and 39 had the outcome "other").

Because listeners were asked to make judgments about the interviewer's personality, contacts had to contain enough speech to make these determinations. The minimum amount of speech required for inclusion was a statement of name and affiliation. Ten contacts were omitted because no speaking turn contained enough introductory interviewer speech for listeners to assign ratings. Another 51 contacts were omitted because the interviewer asked for a particular person by name (indicating that the interviewer had already spoken at length to someone in the household, and the persuasion process was likely to be quite different than in the other contacts), and six more were left out

because the interviewer indicated that the interview had already been started in a previous contact, so the outcome of the invitation was no longer in question.

The 283 contacts ultimately selected for inclusion included 118 (42 percent) with an outcome “agree” and 165 (58 percent) with an outcome of “refuse.”

Listeners were not told the likelihood of either outcome.

Selected contacts were from three studies: the Survey of Consumer Attitudes, or SCA (n=168), the National Study on Medical Decisions, or NSMD (n=110), and the Mississippi Community Study, or MCS (n=5). A total of 49 interviewers were represented in the dataset.

2.2.3 Listeners' Survey: Editing of Contacts

The selected contacts were edited using Praat software for acoustic analysis (<http://www.fon.hum.uva.nl/praat/>). All speech other than the selected interviewer turn was cut from the contact. Typically, the interviewer's first turn was used, although for 15 contacts, a later turn was used, generally because there was not enough uninterrupted interviewer speech in the first turn, or because it was clear to the interviewer that she was speaking to a child (for example, “Is your mom or dad home?”). In 165 contacts, an entire turn was used; in the remaining 118 contacts, the turn was cut short to avoid places where interviewer and answerer were speaking simultaneously; for example, “I'm calling about our study on medical decisions” was cut to “I'm calling about our study” if the answerer started talking during “medical.”

After selecting the interviewer speech to be used, the recording was amplified to use the full range of sounds that a recorded voice would make. Amplification

was maintained at the same level for all contacts, thus making all contacts comparable in volume.

Finally, to preserve interviewers' anonymity, the interviewer's name in each contact was replaced with a quarter-second-long tone. For consistency, this was done even in the few cases where the interviewer only said her first name.

2.2.4 Listeners' Survey: Data Collection

Data collection was conducted by a commercial vendor, Lightspeed Research (<http://www.lightspeedresearch.com/>). Fifteen thousand invitations were sent to members of Lightspeed's own volunteer online panel, and the study was fielded from August 12–18, 2011 among 3,403 panel members, referred to hereafter as "listeners."

Listeners were screened to ensure that they were 18 years of age or older (as would be any eligible respondents to the surveys represented by these contacts), and that they characterized their ability to understand spoken English as "excellent" or "good." This screening criterion was deemed necessary for listeners to be expected to make personality judgments about the interviewer based on brief speech clips. Thirty-nine potential listeners were screened out at this stage: eight who were under age 18, and 31 who did not have the required English abilities.

After their eligibility for the study was determined, listeners were exposed to an "introductory" audio clip and asked to identify one of the words in the clip. The purpose of this exercise was threefold: first, to ensure that listeners were using a computer with working audio; second, to familiarize them with the type of audio

they would be hearing during the survey; and third, as a quality-control check to ensure that listeners could sufficiently distinguish words in the contact. One hundred twenty-six potential listeners were screened out at this stage.

While the mean exposure length of introduction excerpts was 10.32 seconds, the range was wide: from 2.3 to 49.2 seconds. To roughly match the burden on raters and ensure that none received multiple long introductions, excerpts were stratified into five groups based on logical length categories. Each listener received two sets of five introductions, each consisting of one randomly selected excerpt from each length category. For each of the first five introductions, listeners predicted the likelihood that the introduction had resulted in agreement; for each introduction in the second group, none of which overlapped with the first group, they rated the interviewer on the 12 characteristics outlined in Table 2.1.

With the resources available, it was possible to obtain 30 responses for each of the 283 contacts in the corpus. This allowed me to use all available resources while making the best possible use of the large corpus of contacts available. I considered an alternative design whereby a greater number of listeners would hear a smaller number of contacts (to obtain, for example, 100 responses per contact) but recognized that one of the strengths of this project, particularly compared to other studies in this area, was the large number of contacts available.

In order to obtain at least 30 responses per introduction in the likelihood prediction section and at least 30 sets of judgments in the ratings section, while still exposing each listener to one excerpt from each length group for each

section, introductions in the smaller groups were played for more listeners than the introductions in the larger groups.

For example, as Table 2.2 shows, Group 2 was the smallest group, containing 45 contacts. Each one was included in a set with a contact from each of the other, larger groups. Even after 30 ratings were collected for each of the contacts in Group 2, exposure of these contacts continued until a minimum of 30 ratings had been obtained for every contact in the larger groups.

Table 2.2: Audio File Groupings

Group	Length	Number of contacts	Mean number of listeners
1	Under 5 seconds	62	54.4
2	5–6.9 seconds	45	76.0
3	7–9.9 seconds	53	64.2
4	10–14.9 seconds	70	48.6
5	15 seconds or more	53	64.2

As Table 2.3 demonstrates, the distribution of contact lengths was similar across studies, with approximately a quarter of SCA contacts, and a fifth of NSMD and MCS contacts, falling into the shortest group. Between 17 percent and 21 percent of contacts fell into the longest group. NSMD had a greater percentage of contacts in the 10–14.9 second range, while SCA had a greater percentage in the 7–9.9 second range, and MCS had a greater percentage in the 5–6.9 second range.

Table 2.3: Audio File Groupings by Study

Length	SCA	NSMD	MCS
Under 5 seconds	24%	19%	20%
5–6.9 seconds	16%	15%	40%
7–9.9 seconds	23%	13%	0
10–14.9 seconds	20%	33%	20%
15 seconds or more	17%	21%	20%
Total contacts	168	110	5

Table 2.4 shows that the proportion of contacts with each outcome was similar for each length category. There were slightly more refusals in the 7–9.9 second group and slightly more agrees in the 5–6.9 and 10–14.9 second groups, but virtually no difference in the shortest and longest length categories.

Table 2.4: Audio File Groupings by Outcome

Length	Agree	Refuse	Total
Under 5 seconds	23%	24%	22%
5-6.9 seconds	19%	13%	16%
7-9.9 seconds	14%	22%	19%
10-14.9 seconds	27%	23%	25%
15 seconds or more	20%	18%	19%
Total contacts	118	165	283

Listeners were required to play each contact before they were allowed to listen to the next contact. They were given the option to play each contact as many times as they wanted before advancing. On average, listeners listened to audio clips 1.07 times for the likelihood predictions and 1.10 times for the ratings. The number of listens decreased monotonically for each subsequent exposure (e.g., the first clip heard for a given question in the ratings task generated the largest number of listens).

2.2.5 Practitioners' Survey: Questionnaire Development

A questionnaire was developed to assess practitioners' ratings of the importance of various behaviors and attributes to telephone interviewers' success, as well as to illuminate practitioners' current focuses in hiring and training telephone interviewers. This questionnaire was developed based on preliminary qualitative conversations with two employees at the University of

Michigan Survey Research Operations (UM SRO) unit: one a former call center manager and one a Survey Specialist.

Areas of investigation covered in the survey included:

- Size, workload, and clientele of organization;
- Contributors to a telephone interviewer's success;
- Importance of different vocal and personality traits in hiring telephone interviewers; and
- Emphases in telephone interviewer training.

All questions measuring importance used a four-point scale, with a midpoint deliberately excluded to avoid respondents defaulting to this point.

The final survey (see Appendix 4) was programmed in Qualtrics, an online survey tool which the University of Michigan is licensed to use.

2.2.6 Practitioners' Study: Sampling

A sampling frame was developed which included some personal contacts of the investigator, but relied heavily on a list of all members of the Association of Academic Survey Research Organizations (AASRO). For organizations for which a named individual was not known, the investigator looked at the organization's Web site to find a relevant contact. Targeted titles included "Field Supervisor," "Call Center Manager/ Supervisor," and "Operations Manager." For smaller organizations or those with limited staff, the director or assistant director was contacted.

The final sample consisted of 113 individuals at 108 organizations, including 2 government, 92 academic, 3 not-for-profit, and 11 for-profit organizations.

2.2.7 Practitioners' Study: Data Collection and Respondent Overview

An email invitation (see Appendix 5) was sent to all 113 individuals on June 5, 2011 from the investigator's University of Michigan email address.

Two weeks after the initial invitation was sent, a reminder email was sent to all members of the original frame with working email addresses, with the exception of those participants who had already provided their email addresses (respondents were given the option to provide their email addresses if they wished to receive a copy of the results), and those sample members who had requested no further contact.

The survey was completed by 44 respondents before it was closed on July 12, 2011, resulting in a 42 percent response rate.

Respondents represented a wide range of organizations in size, workload, and clientele. The median number of computer-assisted telephone interviewing (CATI) stations in respondent organizations was 25 and the mean was 56, but number of stations ranged from 9 to 450. Close to half (42 percent) of respondents reported that their organization had conducted fewer than 5,000 telephone interviews in 2010, while an equal percentage reported that their organization had conducted 10,000 or more interviews. The remaining 16 percent of respondents reported that their organization had conducted between 5,000 and 10,000 interviews in 2010.

Respondents were not asked to indicate the sector they worked in; however, of the 32 email addresses provided by respondents who wished to receive the

survey results, five included domain names of for-profit organizations, one of a government organization, and 26 of academic organizations.

On average, 80 percent of the work from respondents' organizations was conducted for government, non-profit, or academic organizations, while 15 percent of the work was done for for-profit organizations, and 2 percent for "other" organizations (3 percent of respondents answered "not sure" to this question).

2.3 Results

2.3.1 Listeners' Survey: Respondent Overview

Respondents were evenly divided between males (49 percent) and females (51 percent). One-third (33 percent) were aged 60 or older, while 20 percent were 50–59, 18 percent were 40–49, 17 percent were 30–39, and 12 percent were 18–29. Eighty-eight percent of respondents were white, and 81 percent had at least some college education.

2.3.2 Listeners' Survey: Descriptives

On a six-point scale, the mean ratings for each characteristic ranged from 2.50 to 3.89, with standard errors of these means in the .02 to .04 range. The fact that the mean ratings across contacts for each characteristic spanned at least 1.78 scale points indicates that respondents were able to differentiate among the introductions they heard, and spread their ratings out over the whole scale, rather than consistently default to the middle of the scale.

Table 2.5: Description of Ratings

Characteristic	Overall mean (se)	Minimum	Maximum	Spread
Friendly	3.89 (.02)	2.84	4.62	1.78
Scripted	3.78 (.02)	2.70	4.67	1.97
Professional	3.70 (.03)	1.85	4.73	2.88
Competent	3.67 (.03)	1.90	4.67	2.77
Natural-sounding	3.65 (.02)	2.35	4.47	2.12
Confident	3.62 (.03)	1.77	4.63	2.86
Knowledgeable	3.61 (.03)	2.13	4.75	2.62
Genuine	3.59 (.02)	2.41	4.45	2.04
Pleasant to listen to	3.54 (.03)	2.13	4.49	2.36
Enthusiastic	3.43 (.03)	2.25	4.51	2.26
Uncertain	2.70 (.03)	1.83	4.55	2.72
Irritating	2.50 (.02)	1.73	3.51	1.78

2.3.3 Dimensions of Person Perception: Warmth and Competence

The first hypothesis tested was that contact-level ratings of the nine positive characteristics would cluster into two factors representing the two dimensions of person perception described in the literature: warmth and competence. A factor analysis (shown in Table 2.6) was conducted in an attempt to distill the characteristic ratings into fewer categories.

Due to high correlations between ratings of all positive characteristics (shown in Table 2.7), this factor analysis revealed that only one factor, explaining 86 percent of total variance, had extremely high loadings for all nine positive characteristics. Two characteristics expected to be on a “warmth” dimension, “enthusiastic” and “friendly,” also loaded on a second factor, offering some support for the idea that a “warmth” dimension can be distinguished from a “competence” dimension. When two factors are considered, 94 percent of the variance in ratings is explained.

Note that two undesirable characteristics, “uncertain” and “irritating,” loaded negatively, while “scripted” had an extremely low loading on the first factor.

Table 2.6: Factor Loadings

	Factor 1	Factor 2
Confident	0.96	-0.16
Friendly	0.86	0.37
Professional	0.95	-0.22
Pleasant to listen to	0.94	0.09
Competent	0.97	-0.15
Knowledgeable	0.92	-0.17
Natural-sounding	0.93	0.16
Enthusiastic	0.79	0.31
Genuine	0.94	0.16
Scripted	0.03	-0.59
Uncertain	-0.83	0.27
Irritating	-0.79	0.04

A correlation analysis, shown in Table 2.7, found that ratings for all nine positive characteristics were highly correlated. Of note, the three strongest correlations were found between characteristics hypothesized to make up a “competence” dimension: “confident” and “competent” (.97), “professional” (.94), and “knowledgeable” (.92). Correlations across dimensions were lower, such as the correlation of .66 between “enthusiastic” and “professional” (compared to .88 between “enthusiastic” and “friendly”). While it is clear that all positive characteristics are strongly correlated, the higher correlations between characteristics within one dimension compared to characteristics across dimensions may indicate that listeners do distinguish, to some degree, between the two dimensions of person perception.

Table 2.7: Correlations Between Ratings of Positive Interviewer Characteristics

	Conf	Prof	Pleas	Com	Knowl	Nat	Enthus	Gen	Frien
Confident	1.00								
Professional	0.94	1.00							
Pleasant	0.86	0.88	1.00						
Competent	0.97	0.97	0.88	1.00					
Knowledgeable	0.92	0.92	0.82	0.94	1.00				
Natural	0.84	0.88	0.93	0.88	0.83	1.00			
Enthusiastic	0.78	0.66	0.72	0.73	0.68	0.69	1.00		
Genuine	0.87	0.88	0.91	0.90	0.89	0.93	0.75	1.00	
Friendly	0.77	0.72	0.86	0.76	0.70	0.81	0.88	0.84	1.00

Ratings of “uncertain” and “irritating” were highly correlated with each other (.71), but ratings of “scripted” (the only rating to predict outcome, as will be discussed in Section 2.3.4) were not highly correlated with ratings of any other characteristic. The mean correlation between scripted and positive characteristics was .01.

In summary, the high level of correlation between ratings of all nine positive characteristics indicates that raters make judgments on all of these traits based on a single overall impression of positivity. While there is not unequivocal support for the hypothesis that rated characteristics would cluster into two factors representing the “warmth” and “competence” dimensions, there is some evidence that “friendly” and “enthusiastic” may be perceived as being on a different (but not entirely independent) dimension from characteristics such as “competent,” “confident,” “professional,” and “knowledgeable.”

2.3.4 Judged Likelihood Ratings: Association with Rated Characteristics and Contact Outcome

The first section of the listeners’ study asked judges to predict, on a scale from 1 to 6, the likelihood that a contact resulted in agreement with the survey

request. The mean likelihood rating was 3.36 (se=.04). Contrary to my hypothesis, likelihood ratings did not differ significantly by actual contact outcome: the mean likelihood rating for contacts that resulted in refusal was 3.39 (se=.05), while for those with an outcome of agreement, the mean was 3.32 (se=.07).

While likelihood ratings are not associated with actual contact outcome, these ratings are relatively highly correlated with ratings of all positive characteristics (ranging from .46 to .69), and when likelihood was included in a factor analysis, it loaded highly on a factor with all positive characteristics (which explained 85 percent of variance), as shown in Table 2.8. Since judges rated characteristics for one group of contacts and predicted likelihood for another group, this correlation cannot be attributed to the same judges giving high ratings for both positive characteristics and likelihood.

Table 2.8: Factor Loadings When Likelihood is Included

	Factor 1
Confident	0.96
Friendly	0.85
Professional	0.95
Pleasant to listen to	0.94
Competent	0.97
Knowledgeable	0.93
Natural-sounding	0.92
Enthusiastic	0.79
Genuine	0.94
Scripted	0.05
Uncertain	-0.84
Irritating	-0.79
Likelihood	.64

Twelve separate linear regression models were constructed to predict judges' likelihood ratings on the basis of one rated characteristic. All of these models

controlled for the length of exposure, which, as mentioned, ranged from 2.3 to 49.2 seconds. This variable was included in the models to guard against the possibility that judges assigned different ratings when they heard a longer, compared to a shorter, stimulus. Since the same interviewers conducted multiple contacts (the 283 contacts in the dataset were conducted by 49 interviewers), a term was also added to account for clustering by interviewer. This takes into account any random effects of interviewers. For example, an interviewer with a Southern accent or a lisp may be consistently (over all her contacts) rated as more “friendly” or less “professional.” Since “accent” and “lisp” are not included in the model as variables, these attributes would be captured by the random effects term.

These twelve models revealed that all positive characteristics, and “scripted,” were significant positive predictors of likelihood judgments, while “uncertain” and “irritating” significantly predicted likelihood judgments in a negative direction. As mentioned above, no raters judged the same contacts for likelihood and characteristics; therefore, it can be concluded that ratings of positive characteristics by some judges are predictive of higher likelihood ratings by other judges. At the same time, higher ratings of negative characteristics, such as “uncertain” and “irritating,” predict that other judges will rate the likelihood of success for these contacts lower.

A larger model was constructed to predict judges’ ratings of the likelihood that a contact resulted in agreement. Similar to the individual models described

above, this model controlled for exposure length and accounted for clustering by interviewer. This model included three covariates:

- The contact’s factor score (from the factor analysis described in Table 2.6) for the factor which encompassed the nine positive characteristics but excluded likelihood ratings. The factor score is calculated for each contact by multiplying the mean rating for a characteristic by that characteristic’s factor loading, then summing all products. It can be interpreted as an aggregate measure of all positive characteristics and is more parsimonious than modeling with all nine highly correlated characteristics separately.
- The contact’s mean scriptedness rating, which did not load highly on the factor but was shown in an individual model to be predictive of higher likelihood ratings.
- The interviewer’s length of tenure (in days) at the time of the contact. Longer tenure has been shown in the literature (Singer, Frankel, and Glassman 1983) to be associated with a higher response rate.

As shown in Table 2.9, in this model, both scriptedness and the factor score encompassing all positive characteristics predict higher ratings by judges of the likelihood that a contact would result in agreement. Interviewer experience was not a significant predictor of likelihood ratings.

Table 2.9: Predicting Likelihood Ratings

	Coefficient	SE	Z	P
Scriptedness	.42	.09	4.59	0.000
Factor score	.40	.03	13.40	0.000
Interviewer experience	-.00001	.00003	-0.55	.584
Length of exposure	.02	.004	4.99	0.000

It is clear from this analysis that interviewers who are perceived more positively are also judged (by other raters) as more likely to have success in their contacts. Similarly, more scripted interviewers are expected to be more successful. However, listeners' ratings of the likelihood of agreement were poor predictors of the actual outcome. In a model predicting actual outcome while controlling for length of exposure and accounting for clustering by interviewer, likelihood ratings were not a significant predictor of actual contact outcome. This contradicted my hypothesis that judges' likelihood predictions would accurately predict the actual outcome of a contact.

2.3.5 Characteristic Ratings as Predictors of Actual Contact Outcome

The following hypotheses were tested: that when ratings of nine positive interviewer characteristics (enthusiastic, friendly, natural, genuine, pleasant to listen to, confident, professional, competent, and knowledgeable) were high and ratings of three negative characteristics (irritating, uncertain, and scripted) were low, a contact's actual success would be more likely than when the positive characteristics were rated lower and the negative characteristics were rated higher.

This hypothesis was partially supported. Twelve bivariate models were constructed, using the mean rating on each characteristic to predict agreement and accounting for clustering by interviewers. Of these, only the model for scripted had a significant coefficient ($b=-1.05$, standard error=.40, $z=-2.59$, $p=0.010$), indicating that perceptions of the interviewer as more scripted

decrease the likelihood of a contact's success. These results persisted when the models controlled for the length of exposure (ranging from 2.3 to 49.2 seconds), and also when only the subset of 123 contacts with the longest exposure lengths (at least 10 seconds) were analyzed.

For contacts resulting in agreement, the mean scriptedness rating was 3.72 (se=.03), while for contacts resulting in refusal, the mean rating was 3.82 (se=.02). A t-test showed these means to be significantly different ($t [281]=2.67$, $p=0.004$).

In addition, the factor score from the first factor in the factor analysis described in Table 2.6 (which includes all rated characteristics and is a measure of how positive an interviewer is) was computed and used as a covariate in a multiple logistic regression model. This model predicted contact outcome using the factor score and the contact's mean scriptedness rating, while controlling for exposure length and interviewer experience and accounting for clustering by interviewer. Only scriptedness was a significant predictor in this model ($t=-2.53$, $p=0.012$); the factor score was not.

Interviewer experience at the time of the contact ranged from 18 days of performing this type of work to 8,324 days, with a median value of 309 days. Interviewer experience significantly predicted ratings of scriptedness when controlling for the length of exposure and accounting for clustering by interviewer ($b=-.000035$, $se=.00002$, $z=-2.21$, $p=0.0127$), indicating that as interviewers gain more experience, they become less scripted. This may be a contributor to the

greater success rates of more experienced interviewers, which have been documented in the literature (e.g., Singer, Frankel, and Glassman 1983).

In summary, there was no support for the hypothesis that positive characteristics would predict a successful outcome. Of all characteristics rated, only scriptedness was associated with success, with less scripted interviewers being more successful than those who were rated as more scripted. The association between scriptedness and outcome held when interviewer experience (which is negatively associated with scriptedness) was controlled for. As a precaution against a type one error due to multiple testing effects, the rigorous Bonferroni correction was applied to this result, leading to the disappearance of the relationship between scriptedness and contact success. While this result is disappointing, I would argue that the independence of scriptedness from all other characteristics tested (as shown in the factor analysis in Table 2.6, where scriptedness has an extremely low loading compare to all the other, highly correlated characteristics) does not in fact necessitate a Bonferroni correction.

2.3.6 Importance of First Impressions: Comparison Between Listeners' and Practitioners' Surveys

The pattern that has emerged, showing a relationship between ratings of an interviewer's scriptedness and contact outcome (but no relationship between other ratings and outcome), may be at odds with widely held assumptions in the research industry as to which interviewer attributes lead to greater success in obtaining interviews. To explore this, I conducted a survey of individuals in

survey organizations who were responsible for hiring and/or training telephone interviewers.

Before the listeners' study was conducted, an online survey was sent to survey practitioners to collect their views on important traits and skills of telephone interviewers. The results show that the practitioners' opinions aligned with the characteristics which predicted higher likelihood ratings (such as competence and confidence), but pointed to a gap between what survey practitioners believe contributes to interviewer success and which ratings predict actual success.

Regardless of whether they were responsible for interviewer hiring, training, or both, all respondents to the practitioners' survey were asked to rate 18 attributes on their importance to an interviewer's response rate. A four-point importance scale was used, from "not at all" important to "extremely" important. (A mean score of 4 indicates that all respondents rated this attribute as "extremely" important.) It is not known whether practitioners based their ratings on the importance of these elements to an interviewer's overall performance, as the question intended to measure, or to her performance in converting the most resistant answerers.

Of the 18 elements rated, the one judged most important by survey practitioners was "the initial impression an interviewer gives to sample members." This contrasts sharply with the finding in the listeners' survey that, aside from scriptedness, no ratings of interviewer characteristics based on early impressions

can predict success on a given contact. However, these ratings do predict judgments of the contact’s likelihood of success.

While results from the listeners’ study indicated that impressions of scriptedness are, in fact, important to the success of a contact (with lower ratings of scriptedness found in successful contacts), just 29 percent of practitioners rated “an interviewer’s ability to ‘ad lib’ or deviate from a script during an introduction” as “extremely important” to that interviewer’s success. Far more important, according to practitioners, were traits such as competence, professional demeanor, and confidence—ratings of which were predictive of listeners’ judgments about the likelihood of success, but not of contact-level outcome. Practitioners viewed characteristics on a “warmth” dimension, such as friendly and enthusiastic, as less important to an interviewer’s success than traits along a competence dimension.

Table 2.10: Practitioners’ Ratings of Importance to Interviewer’s Success

		Mean (sd)	% Extremely Important
1	The initial impression an interviewer gives to sample members.	3.88 (.33)	88%
2	The ability to address concerns expressed by potential respondents.	3.84 (.38)	83%
3	How competent the interviewer sounds to potential respondents.	3.84 (.38)	83%
4	Professional demeanor when talking to potential respondents.	3.81 (.44)	86%
5	How confident the interviewer sounds to potential respondents.	3.81 (.45)	83%
6	The ability to convey knowledge about the study.	3.72 (.46)	71%
7	How genuine the interviewer sounds to potential respondents.	3.70 (.47)	69%
8	An interviewer’s voice that does not sound monotonous (has pitch variability).	3.58 (.5)	57%

9	How friendly the interviewer sounds to potential respondents.	3.53 (.55)	57%
10	The interviewer's speech rate.	3.42 (.59)	48%
11	How enthusiastic the interviewer sounds to potential respondents.	3.42 (.63)	50%
12	A pleasant-sounding voice.	3.41 (.59)	45%
13	The interviewer's ability to follow a script during an introduction.	3.30 (.78)	48%
14	The interviewer speaks without any "um's" or "uh's."	3.16 (.71)	36%
15	The interviewer's ability to "ad lib" or deviate from a script during an introduction.	2.95 (.89)	29%
16	How high or low the interviewer's voice sounds (pitch).	2.70 (.74)	12%
17	The interviewer emphasizes the length of the survey.	2.67 (.67)	12%
18	The interviewer emphasizes the incentive.	2.47 (.74)	10%

Further, among practitioners responsible for training telephone interviewers, just 15 percent (the lowest overall percentage) report that "developing a personalized or non-scripted introduction" is a primary focus of their organization's interviewer training, while 44 percent reported that it is not a focus at all. "Following introductory scripts," by contrast, was a primary training focus for 78 percent of trainers surveyed. This aligns with practitioners' assertion that an interviewer's ability to "ad lib," or deviate from a script during an introduction, is relatively unimportant to the interviewer's overall success, but contrasts with literature finding that interviewers who are allowed to adapt their script have greater success (Houtkoop-Steenstra and van den Bergh 2000; Morton-Williams 1993); it also contradicts the finding in the listeners' survey that higher ratings of scriptedness predict less success at the contact level.

Table 2.11: Emphases in Interviewer Training

		Mean (sd)	% primary focus
1	Administering interviews.	2.98 (.16)	98%
2	Importance of data quality.	2.98 (.16)	98%
3	Standardized interviewing skills.	2.90 (.30)	90%
4	Obtaining interviews in general.	2.80 (.46)	83%
5	Use of CATI system.	2.76 (.42)	78%

6	Following introductory scripts.	2.76 (.53)	78%
7	Responding appropriately to arguments or concerns expressed by potential respondents.	2.76 (.49)	78%
8	Being able to answer questions about the study or organization.	2.68 (.53)	70%
9	Presenting a confident demeanor.	2.61 (.59)	65%
10	Developing knowledge about the study.	2.49 (.6)	53%
11	Obtaining interviews from reluctant respondents (including refusal conversion).	2.41 (.55)	45%
12	Recruiting respondents.	2.35 (.71)	49%
13	Developing a personalized or non-scripted introduction.	1.73 (.72)	15%

The results discussed in this section demonstrate a disconnect between listeners and practitioners. While the listeners' data indicate that no characteristics other than scriptedness are necessary to a contact's success, practitioners place less emphasis on reducing scriptedness, and more on other impressions conveyed by interviewers.

2.3.7 Relationship Between Vocal Characteristics, Ratings, and Contact Outcome

Two hypotheses were tested involving interviewers' vocal characteristics. It was hypothesized that vocal characteristics shown in the literature (Oksenberg, Coleman, and Canell 1986; van der Vaart et al. 2005) to be associated with competence—such as a moderately fast rate of speech and lower voice (measured by fundamental frequency)—would predict higher ratings of characteristics on this dimension (competent, confident, knowledgeable, and professional), while vocal characteristics associated with warmth—such as a higher fundamental frequency—would predict higher ratings of characteristics on

a “warmth” dimension, including enthusiastic, friendly, genuine, natural, and pleasant to listen to.

While some vocal characteristics are predictive of ratings, these hypotheses were not supported in my dataset. A faster speech rate is associated with higher ratings on all positive characteristics. Faster speech predicts, as hypothesized, higher ratings of competent, confident, knowledgeable, and professional. Faster speech also predicts higher ratings of “warmth” characteristics such as enthusiastic, friendly, genuine, natural, and pleasant to listen to, although the coefficients for these variables are slightly lower than those in the models predicting traits on a competence dimension. Table 2.12 describes 11 separate models, one predicting each characteristic. The significance of the coefficients for speech rate in all of the models outlined in the below table remain even when a Bonferroni correction is applied.

Table 2.12: Predicting Ratings with Interviewer Speech Rate (controlling for exposure length and accounting for clustering by interviewer)

Dependent Variable	Coefficient	SE	Z	P
Confident	.29	.06	5.25	0.000
Professional	.26	.05	4.80	0.000
Competent	.25	.05	4.90	0.000
Knowledgeable	.26	.05	5.34	0.000
Friendly	.16	.04	4.18	0.000
Pleasant to listen to	.22	.05	4.82	0.000
Natural	.21	.04	5.11	0.000
Enthusiastic	.23	.05	4.75	0.000
Genuine	.20	.04	5.34	0.000
Irritating	-.13	.04	-3.59	0.000
Uncertain	-.22	.05	-4.28	0.000

As shown in Table 2.13 below, in general, lower voices are associated with desirable characteristics and higher voices are associated with undesirable

characteristics. One exception to this pattern is the desirable trait “enthusiastic”: interviewers with higher voices were rated as more enthusiastic. Application of a Bonferroni correction resulted in significant coefficients for fundamental frequency as a predictor in six models: those predicting professional, knowledgeable, pleasant to listen to, natural, irritating, and enthusiastic.

Table 2.13: Predicting Ratings with Interviewer Fundamental Frequency (controlling for exposure length and accounting for clustering by interviewer)

Dependent variable	Coefficient	SE	Z	P
Confident	-.002	.001	-1.99	0.047
Professional	-.004	.001	-3.88	0.000
Competent	-.003	.001	-2.74	0.006
Knowledgeable	-.003	.001	-2.89	0.004
Scripted	-.001	.001	-1.94	0.05
Pleasant to listen to	-.003	.001	-3.03	0.002
Natural	-.003	.001	-3.19	0.001
Genuine	-.002	.001	-2.11	0.035
Irritating	.003	.001	4.15	0.000
Uncertain	.002	.001	2.12	0.034
Enthusiastic	.003	.001	3.05	0.002

In summary, the hypothesis that vocal characteristics would have different relationships with traits on a “warmth” versus a “competence” dimension is not supported. However, this analysis does show that lower, faster voices are more appealing than higher and slower voices.

Next, a model using interviewer speech rate and fundamental frequency in the clip to which listeners were exposed (controlling for length of exposure and accounting for clustering by interviewer) was constructed to predict likelihood ratings. An interviewer’s speech rate was positively associated with higher likelihood ratings ($z=3.09$, $p=0.002$). Similarly, a lower fundamental frequency

was a marginally significant predictor of higher likelihood ratings ($z=-1.75$, $p=0.08$). However, neither rate nor fundamental frequency in the exposed clip is significant in a logistic model predicting contact outcome, which controlled for length of exposure and accounted for clustering by interviewer.

Importantly, the vocal characteristics discussed here are only measured for the (often very short) clips of interviewer speech tested in the Web survey. While work by Benkí et al. (2011) finds higher rates of agreement when the interviewer speaks at a rate of approximately 3.5 words per second over the course of the introduction, this finding does not hold in the current study, where only the rate in the brief introduction excerpt comprising the exposed clip was measured.

2.3.8 Vocal Characteristics: Contrast with Practitioners' Study

In the practitioners' survey, those who indicated that they were responsible for hiring telephone interviewers were asked to rate the importance of 12 criteria (again from not very important to extremely important) in their hiring decisions. A total of 40 individuals responded to this battery.

English language fluency, trainability, and reliability were the most important hiring criteria, with speaking voice ranked fifth, indicating that it carries moderate importance in hiring decisions—more so than prior experience or references from previous employers. This is in surprising contrast to the practitioners' view, shown in Table 2.10, that “a pleasant speaking voice” is perceived to be among the least important contributors to an interviewer's response rate.

Table 2.14: Practitioners' Emphases in Hiring Telephone Interviewers

		Mean (sd)	% Extremely Important
1	Candidate seems trainable.	3.83 (.38)	85%
2	English language fluency.	3.83 (.43)	88%
3	Candidate seems reliable.	3.80 (.41)	80%
4	Literacy level.	3.78 (.42)	78%
5	Speaking voice.	3.71 (.45)	73%
6	Beliefs about how the individual will interact with respondents and potential respondents.	3.56 (.50)	58%
7	Performance in training exercises.	3.43 (.55)	43%
8	Beliefs about how the individual will get along with other team members.	2.88 (.67)	15%
9	Ability to multi-task.	2.73 (.84)	18%
10	References from previous employers.	2.51 (.78)	13%
11	Professional appearance.	2.46 (.75)	8%
12	Prior experience.	2.37 (.81)	10%

Practitioners who indicated that a candidate's speaking voice was "somewhat" or "extremely" important to their hiring decisions were asked to rate the importance of nine attributes of voice in their hiring decisions. Here, the impressions conveyed by the voice (professionalism, confidence, enthusiasm, and knowledge) were considered much more important than vocal characteristics such as rate, volume, and pitch.

Table 2.15: Importance of Vocal Attributes in Hiring Decisions

		Mean	% Extremely important
1	Voice conveys confidence.	3.63 (.59)	68%
2	Voice sounds professional.	3.61 (.59)	65%
3	Voice sounds friendly.	3.59 (.55)	60%
4	Voice conveys knowledge.	3.56 (.6)	60%
5	Voice conveys enthusiasm.	3.46 (.64)	53%
6	Voice is pleasant to listen to.	3.32 (.65)	40%
7	How slow or fast the individual speaks (speech rate).	3.20 (.72)	35%
8	How soft or loud the voice sounds (volume).	3.00 (.6)	15%
9	How high or low the voice sounds (pitch).	2.76 (.71)	10%

Practitioners' low ratings of the importance of vocal characteristics such as rate and pitch indicate that they may not be aware that these characteristics actually drive perceptions of characteristics which they deem as important, such as sounding professional, confident, and enthusiastic.

2.4 Conclusions

This exploratory research has found that survey practitioners believe firmly that initial impressions of an interviewer are critical to that interviewer's success. These impressions, measured by judges' ratings of interviewer characteristics in a short clip of an early speaking turn by an interviewer in a telephone survey introduction, align with ratings by different judges, again based on this brief clip, of whether or not the contact resulted in success.

However, despite practitioners' belief that these traits are key to success, most ratings of interviewer traits such as competence, confidence, and professionalism based on this brief exposure are not predictive of the ultimate outcome of the conversation. One exception to this is ratings of scriptedness, which decrease with interviewer experience and are significant predictors of contact outcome. This can be an important finding for survey practice, as an emphasis on decreasing the scripted or "robotic" nature of survey introductions may well serve to increase response rates. Currently, practitioners do not recognize scriptedness as detracting from an interviewer's success. In fact, practitioners widely believe that an interviewer's ability to follow a script is far more important than the ability to "ad lib" during an introduction—precisely the opposite of what the contact-level ratings suggest.

The salience of scriptedness compared to other attributes may be due to the fact that scriptedness is easiest to discern in such a brief exposure, while other attributes require longer exposure to become apparent. A study which exposes listeners to longer excerpts of interviewer speech may obtain different results.

It should also be noted that the difference in the importance of scriptedness between listeners (who considered it very important to success) and practitioners (for whom this was not as important) may be due to different interpretations of the word “scripted.” Listeners may have taken this term to mean that an interviewer sounded more robotic or like she was reading from a script in the brief excerpt, while practitioners may have been thinking about scriptedness over the life of a contact. A more “apples to apples” comparison between these two groups, using different question wording, is necessary to further solidify this conclusion.

It is important to note that the contacts selected came from a “shared sample” environment, where cases, or households, requiring multiple calls are called back by multiple interviewers, in contrast to assigning one interviewer to continually follow up with the same household. In such an environment, some interviewers who are expert “refusal converters” are assigned difficult cases, where answerers have repeatedly refused to participate. Such an interviewer may be perceived as having many positive characteristics, but still have calls which do not result in agreement, because of the nature of answerers being called.

Additionally, there is the issue of “hard-core nonrespondents” who will not respond to a survey regardless of what the interviewer says or how she sounds.

While these analyses attempted to exclude such individuals by eliminating invitations in which the answerer hangs up during or directly after the interviewer's first speaking turn, it may be that a higher threshold is needed to truly rule them out. Some hard-core nonrespondents may stay on the phone slightly longer, but still have no intention of being persuaded to respond.

The ability of listeners to tease apart the classic person perception dimensions of "warmth" and "competence" based on brief exposures to an interviewer's speech remains an open question. All ratings of positive characteristics are highly intercorrelated, indicating that these two dimensions are correlated with each other and may not be easily discerned. However, slightly higher correlations among ratings on a "competence" dimension compared to correlations of characteristics across dimensions, coupled with the finding from a factor analysis that "enthusiastic" and "friendly" were the only characteristics to load on a second factor, indicate that these dimensions may in fact be differentiated in this setting. The finding that practitioners consider these dimensions differentially important to an interviewer's success further indicates that it may be possible to tease apart "warmth" and "competence."

Finally, vocal characteristics such as rate and fundamental frequency, while not considered by practitioners to be of tremendous importance in relation to an interviewer's success, are positively associated with not only ratings of personality characteristics, but also with judges' predictions of the likelihood of a contact's success. Interviewers with lower voices and those who spoke faster were perceived more favorably, and their contacts were judged as more likely to

succeed, compared to their slower speaking and more high-pitched counterparts. Since rate in particular is trainable, this finding could provide a basis for the recommendation that survey practitioners instruct interviewers to aim for an optimal rate of speech (3.5 words per second according to Benkí et al. 2011) in their introductions.

2.5 Limitations and Suggestions for Future Research

This study was not without limitations. First, the raters were not actual respondents; rather, they heard recordings of the interviewers' voices via a Web survey instead of on the phone. Further, Web survey panel members who served as raters did not match the United States population in terms of education (81 percent of raters had at least some college education, compared to 55 percent of the national population, according to the 2010 US Census). Having "real" respondents to a telephone survey rate an interviewer on these characteristics (perhaps via a Web survey immediately after hearing a survey introduction on the telephone) could lead to results that more accurately show the relationship between perceptions of an interviewer and contact outcome. This idea will be discussed further in Section 4.4.8.

Also, because recordings from actual contacts were used, the length of exposure varied and was often as short as 2.3 seconds. While there did not seem to be any differences in results based on exposure length, it is possible that exposing respondents to only longer excerpts may change the results. For example, in short exposures, scriptedness may simply be more salient than other characteristics, but this difference may disappear if longer contacts are used.

While scriptedness was a significant predictor of outcome in models controlling for exposure length and accounting for random interviewer effects, the possibility exists that the significance of this variable may be due to multiple testing effects; when a Bonferroni correction was applied, scriptedness was no longer significant in predicting outcome. If the effect of scriptedness can be replicated, it can be stated with greater certainty that reducing perceived scriptedness in survey invitations impacts an interviewer's success.

Finally, analyses at the contact level, incorporating only one side of the conversation, may be unreliable and marred by noise. An interviewer could receive high ratings on all the "right" characteristics, but still be met with rejection if a "hard-core" nonrespondent is reached; conversely, an interviewer with low ratings may still have success on individual contacts with answerers who are less discerning about the impressions given by their conversational partners. Interviewer-level analyses could provide greater stability than contact-level predictions. Given the small number of contacts per interviewer in this dataset (a mean of 10.4 contacts per interviewer, with only 7 interviewers having 10 or more contacts), such analyses were not conducted.

A larger study, collecting ratings of characteristics for a larger number of contacts per interviewer to measure the impact of ratings on overall success rates, is recommended.

Chapter 3: Interviewer Responsiveness

3.1 Introduction

An initial impression of an interviewer as less scripted may be enough to get his or her foot in the door, so to speak, by keeping an answerer on the phone for a few additional seconds. However, this first impression in and of itself is likely not to be enough to guarantee cooperation with the survey request. In this context, interviewers need to be responsive to answerers throughout the contact.

Survey introduction contacts (defined as consisting of the exchange from the first “hello” until either the first question of the interview or until the answerer hangs up) are a unique type of conversation. In most conversations, the participants are working together toward a common goal (Clark and Schaefer 1989). However, in this instance, they are often oriented toward opposing goals: the interviewer wants to complete the survey, while the answerer wants to get off the phone as quickly as possible. I posit that responsiveness on the part of the interviewer can help an interviewer convert an answerer to the goal of survey completion.

The exploratory study described in this chapter will address the following questions:

- Are more concerns expressed by answerers who ultimately agree to, refuse, or defer from survey completion?

- Do the types of concerns that are most frequently expressed by answerers differ whether they ultimately agree to, refuse, or defer from survey completion?
- Are answerers more likely to use conversation starters such as questions or topics not directly related to the task at hand, which give interviewers an opportunity to tailor or personalize their introductions, if they ultimately agree, refuse, or defer?
- Are interviewers more responsive to concerns and conversation starters in contacts that result in agreement, refusal, or deferral?
- Are interviewers more or less responsive to different types of concerns (for example, concerns about length versus statements of disinterest)?
- Are interviewers consistent in their levels of responsiveness, or does responsiveness vary for each interviewer between contacts?
- Does the association between interviewer scriptedness and ultimate contact outcome, found in Chapter 2, persist when responsiveness is included in the models?

I hypothesize that interviewers who are more responsive to answerer concerns and conversation starters will be more successful in individual invitations than interviewers who are not. At the same time, I hypothesize that answerers who are destined for agreement will provide interviewers with more conversation starters, in the form of peripheral conversation and questions, than those who ultimately refuse, who will express more concerns. Finally, I hypothesize that concerns expressed by answerers to calls which result in

scheduled callbacks will be predominantly related to the timing of the call, while answerers who refuse will be more likely to express concerns such as “I’m not interested.”

The study described in this chapter used a detailed coding scheme to count and classify opportunities posed by answerers for interviewer responsiveness. The prevalence of different types of opportunities (both concerns and conversation starters) across different outcomes (agree, refuse, scheduled callback) is discussed. This is followed by an explanation of the construction of an interviewer responsiveness score, where points are added or subtracted based on the interviewer’s responses to concerns or conversation starters presented by the answerer, and the presentation of differing responsiveness scores for contacts with different outcomes. Interviewer techniques for addressing particular concerns are also explored. The final two subsections of results connect this work to the studies described in Chapter 2, by presenting the results of models predicting contact-level success based on both listeners’ first impressions and interviewer responsiveness, and discussing the responsiveness results in light of the practitioners’ study detailed in Chapter 2. This chapter concludes with practical implications and suggestions for future research.

3.2 Data and Methods

The corpus of 1,380 contacts described in Chapter 2 was the basis for contacts included in this study. All of these contacts were transcribed by speech-language pathology students at Michigan State University. They transcribed the interactions at the turn level, where a turn is defined as all the speech by one

speaker that occurs before the other speaker begins to talk. Transcripts used standardized notation to indicate repairs (such as “Mon-Sunday”), restarts (“Mon-Monday”), and rising intonation. The following example transcript shows two answerer and two interviewer turns:

Answerer: Hello?

Interviewer: Hello, this is Jane Doe from the University of Michigan.

Answerer: Ok?

Interviewer: Um, you may have gotten a letter from us recently, um, about a chance for you or someone in your family to earn twenty dollars for participating in a research study about the economy.

Following transcription, seven undergraduate and graduate (master’s level) students in the humanities and social sciences at the University of Michigan coded the content and paralinguistic behavior in each transcript. Prior to coding the interaction, coders decomposed each conversational turn into one or more moves. Good reliability was achieved on both the decomposition and coding portions of this task (see Appendix 6). A list of 61 mutually exclusive move descriptions was created, including “incentive-related question” (householder), “not enough time” (householder), “offer to break interview into pieces” (interviewer), “description of survey” (interviewer), and “backchannel” (interviewer or answerer); see Appendix 7 for a complete list of move codes. The following hypothetical contact illustrates the kinds of codes (*in parentheses*) that were assigned to each move (segment of the transcripts beginning with “Answerer:” or “Interviewer:”).

Answerer: Hello? (*formal greeting*)

Interviewer: Hello, (*formal greeting*)

Interviewer: this is Jane Doe from the University of Michigan. (*self-identification*)

Answerer: Ok? (*backchannel*)

Interviewer: Um, you may have gotten a letter from us recently (*follow-up statement*)

Interviewer: about a chance for you or someone in your family to earn twenty dollars for participating in a research study (*incentive related comment*)

Interviewer: about the economy. (*description of survey*)

A subset of these transcribed, coded contacts was used for the study described in this chapter. As in the listeners' study described in Chapter 2, contacts were excluded if they were conducted by male interviewers; were conducted as part of the Gujarati Community Study (a study conducted among members of a particular linguistic and cultural group); or involved the answerer hanging up during or immediately following the interviewer's first turn. A total of 674 contacts met the inclusion criteria; of these, 48 with outcome "hang up" or "other" were excluded, leaving 626 contacts to be analyzed in this chapter. Of these, 157 had the outcome "agree," 194 had the outcome "refuse," and 275 had the outcome "scheduled callback," in which an interviewer stated that she or a colleague would call the household again. Contacts were selected from four studies: Interests of the General Public (n=45); Mississippi Community Study (n=12); National Study on Medical Decisions (n=208); and the Survey of

Consumer Attitudes (n=361). A total of 58 interviewers were represented in this dataset.

Additional coding was done on the 626 contacts in this subset to capture two types of tailoring opportunities expressed by answerers: concerns and conversation starters. A coding scheme (see Appendix 8) was developed to classify 13 specific concerns expressed by answerers and responses to the concerns by interviewers. Common concerns included the timing of the call, the length of the survey, an inability to hear the interviewer clearly, or lack of interest in participation. Coding of interviewer moves indicated if the interviewer addressed a concern with a relevant response.

The examples below (from actual contacts) show (1) a successful interviewer attempt at responsiveness, and (2) an irrelevant interviewer response to the answerer's concern.

1) Answerer: There's a lot of questions that we probably couldn't even answer.

Interviewer: Well, it's not a test or anything.

2) Answerer: There's only two of us and my husband's in the shower and I'm in the middle of making dinner.

Interviewer: Well, this is a very important study.

Concerns may also be related not to the larger survey request but to characteristics of the call, for example, "I can't hear you," "You're talking too fast," or "I'm in a hurry, can we do this quickly?" Since both transcripts and audio recordings were used, in these instances, coders were able to assess

interviewers' responsiveness not only by what they said (for example, "I'll fix the headset. Can you hear me now?"), but also by how they said it, that is, whether they adjusted their speaking volume or speed. Interviewers who acknowledged a concern (for example, "I understand" or "Oh, you're busy now?") were not considered to have addressed or offered a solution to a concern.

Concerns can be expressed either as questions, statements, or commands, as shown in the following examples:

- 1) Answerer: How long is it going to take?
- 2) Answerer: Uh, excuse me, I do not speak, uh, really much English.
- 3) Answerer: Listen, please don't call me. I'm on a Do Not Call list.

Answerers may also present what I will refer to as "conversation starters," which are moves that are not concerns, but that provide opportunities for interviewers to give a response that demonstrates their attentiveness. Such remarks could be phrased as either comments or questions; they could include observations about the survey task, or they could be peripheral to the task—the point is that the interviewer can respond in a way that shows she has understood and thought about the answerer's comment. As in any conversation, the interviewer can "take the bait" and acknowledge these remarks with a relevant response—what Clark and Schaefer (1989) refer to as "contributing to discourse." Interviewers can also fail to effectively respond to answerers' conversation starters, either by offering content that is not relevant to what the answerer has said, or by saying something with no substance, as in example (2) below. The examples below show a relevant interviewer response to an

answerer's conversation starter (1) and a missed opportunity to respond to a conversation starter (2).

1) Answerer: You can just put that everything is way out of control. I work twelve hours a day to pay for everything.

Interviewer: Well, that's why we do want to get your opinions. It is very important.

2) Answerer: The economy? I don't like it and that's the end of it.

Interviewer: Oh. Well.

Transcripts were coded and analyzed in Sequence Viewer, which is software designed specifically to analyze relationships between sequential events. Using this software allows for analysis not only as to whether one event follows another (for example, is an answerer's concern about the survey length followed by the interviewer addressing this concern?), but also the lag (in speaking turns) between events (for example, does the interviewer address the answerer's concern immediately, or are there several intervening turns?). Coders viewed each contact as a whole and indicated when concerns were addressed or conversation starters responded to later in a contact. Lag is important for the analyses in which responding to concerns or conversation starters immediately after they are uttered (as in the first example below) is given more weight than doing so later within the exchange (as in the second example below).

1) Answerer: I'm sorry dear, I'm on my way out the door.

Interviewer: Ok, we'll call back.

2) Answerer: We're just about to have dinner.

Interviewer: Oh, you are?

Answerer: Uh huh.

Interviewer: When would be a better time to reach you?

As the Sequence Viewer illustration in Figure 3.1 shows, codes can be assigned at the level of individual moves (for example, the presence of a specific concern, indicated by the solid circle), as well as at the level of the contact (for example, the total number of concerns, indicated by the dashed circle).

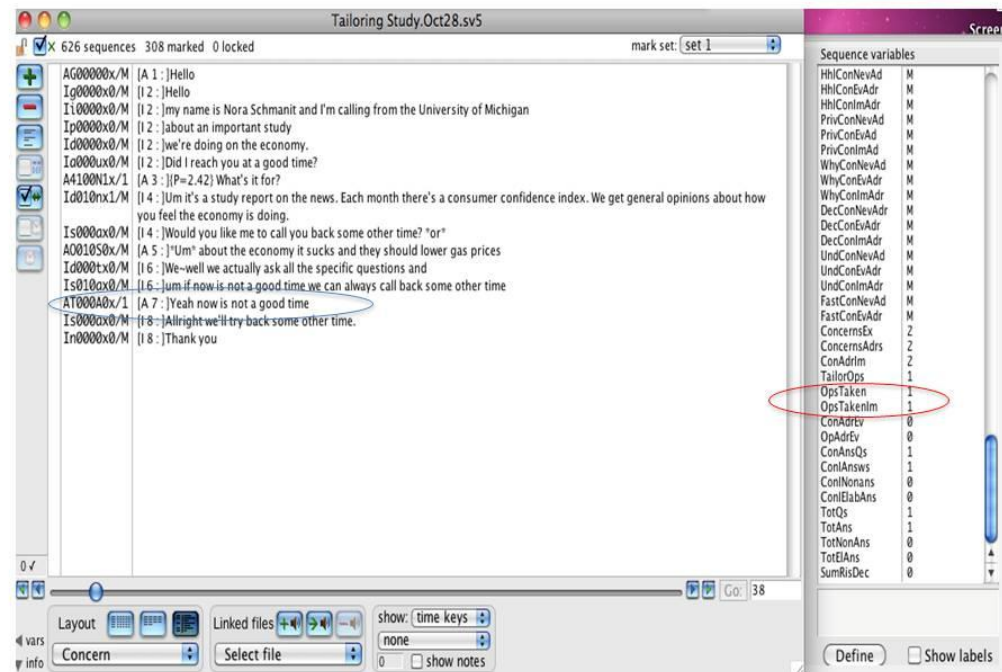


Figure 3.1: Sequence Viewer Screen Shot

Coding was done by a research assistant who had recently completed a bachelor's degree at the University of Michigan. He had worked as a coder on the original project, described in Section 2.2.2, for which the corpus was created, and he was therefore familiar with both the structure of the dataset and with Sequence Viewer software. Weekly meetings between the coder and investigator

allowed for discussion and resolution of questions about the coding scheme, as well as the ongoing adaptation of the coding scheme. Of the 626 contacts used in this study, 200 randomly selected contacts (stratified by outcome) were double-coded by the investigator and coder. Inter-coder reliability was assessed using Cohen's kappa, which ranges from zero to one; Landis and Koch (1977) characterized kappa values of 0–.20 as slight, .21–.40 as fair, .41–.60 as moderate, .61–.80 as substantial, and .81–1 as almost perfect agreement.

Using Landis and Koch's characterizations, there was substantial inter-coder reliability (.689) for the variable capturing expression of concerns, conversation starters, and responses. This variable had 30 possible values: 15 options for answerer turns (turns could be one of 13 concerns, a conversation starter, or none of these) and 15 for interviewer turns (a response to one of 13 concerns, a response to a conversation starter, or none of these). Coders agreed on the assignment of this code in 80 percent of moves.

There was substantial inter-coder reliability (.674) as to whether a move contained a conversation starter, and moderate reliability (.485) on coding whether an interviewer move was a response to a conversation starter. Reliability between coders as to whether an answerer move contained a concern was assessed by recoding all 13 "concern" codes into one value and computing kappa; the value of kappa for this code, .840, was in the range considered "nearly perfect." Similarly, all 13 "response to concern" codes were collapsed into one value and a "substantial" kappa (.751) was computed. Values of kappa for specific and "umbrella" codes are shown in Table 3.1.

Table 3.1: Inter-coder Reliability Measures

	Number of moves in this category as judged by the primary coder	Kappa
Answerer move is a conversation starter (as opposed to concern or neither of these).	449	.674
Interviewer move is a response to conversation starter (as opposed to a response to a concern or neither of these).	226	.485
Answerer move is a concern (as opposed to a conversation starter or neither of these).	674	.840
Interviewer move is a response to a concern (as opposed to a response to a conversation starter or neither of these).	884	.751
Overall assignment of variable capturing conversation starter/ specific concern/ response/ no conversation starter or response by interviewer or answerer.	6,817	.689

Upon completion of coding, an interviewer responsiveness score was computed for each contact. This summary measure takes into account the following elements:

- Concerns addressed within one turn (CA1): The number of concerns expressed by an answerer that are addressed by an interviewer in the interviewer's next speaking turn. I considered this the most responsive way for interviewers to address concerns, and this ideal receives "full credit" in the overall score.

- Concerns addressed later (CAL): The number of concerns expressed by an answerer that are addressed by an interviewer later in the contact, but not in the interviewer's next turn. This still can be considered responsive, but not to the same extent as the previous code; it therefore is given less weight in the calculation of the overall score, earning an interviewer half a point (whereas an immediate response earns one point).
- Concerns never addressed (CNA): The number of concerns expressed by an answerer that are never addressed by the interviewer. This is considered detrimental to an interviewer's success and is assigned a score of -1.
- Conversation starters addressed within one turn (CSA1): The number of conversation starters as described above made by an answerer that are responded to by an interviewer in the interviewer's next turn. Just like concerns, immediately addressed conversation starters earn the interviewer one point.
- Conversation starters addressed later (CSAL): The number of conversation starters by an answerer that are responded to by an interviewer later in the contact but not in the interviewer's next turn. As in the concern calculations detailed above, conversation starters addressed later in the contact are worth half a point.
- Conversation starters never addressed (CSNA): The number of conversation starters made by an answerer that are never addressed by the interviewer. As with concerns, showing a lack of responsiveness by

not addressing these conversation starters is considered detrimental to the success of a contact; therefore, for each instance of a conversation starter never addressed, an interviewer's score is "docked" one point.

- The total number of chances for interviewer responsiveness, calculated as the sum of all concerns and conversation starters:

$$CA1+CAL+CAN+CSA1+CSAL+CSNA.$$

Each contact's interviewer responsiveness score, therefore, is the sum of responses to all concerns and conversation starters over the total number of concerns and conversation starters. The score is computed as follows:

$$[(CA1) + (CAL / 2) - (CNA) + (CSA1) + (CSAL/2) - (CSNA)] / [\text{Total number of concerns} + \text{Total number of conversation starters}].$$

Figure 3.2 shows a plot of decomposed interviewer responsiveness scores for all contacts in this study. The numerator, interviewer responsive behavior, is computed as $[(CA1) + (CAL / 2) - (CNA) + (CSA1) + (CSAL/2) - (CSNA)]$ and is shown on the x-axis, while the total number of opportunities (concerns + conversation starters) is on the y-axis. This plot shows that, in general, interviewers are able to keep pace with the tailoring opportunities presented by answerers; the numerator of responsiveness scores increases in tandem with the number of opportunities. A handful of outliers show the opposite pattern, with low responsiveness to a high number of opportunities.

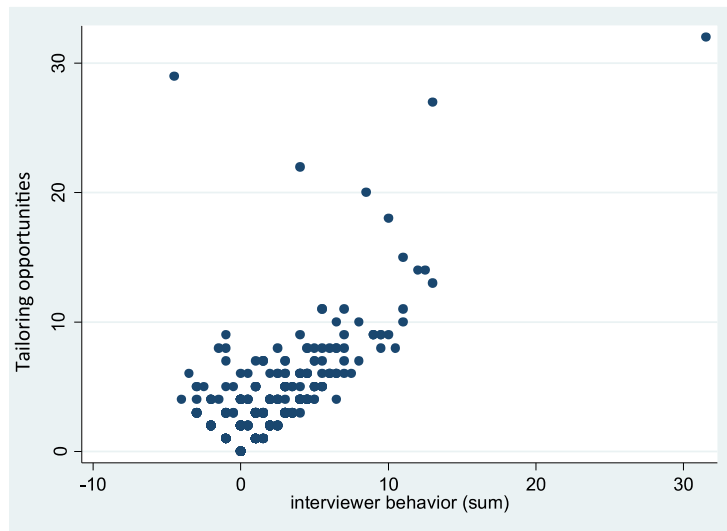


Figure 3.2: Interviewer Responsive Behavior against Tailoring Opportunities

The structure of the responsiveness score assumes that never responding to a concern or conversation starter is detrimental to an interviewer’s success, while an immediate response is the most effective. Addressing a concern or conversation starter after some delay is assumed to be somewhat effective, and half a point was determined to be the fairest value to assign to this type of response; giving no credit for addressing a concern later would not acknowledge the potential benefits of eventually addressing the concern, but a later response is presumably not as helpful as an immediate response and so should not be assigned as large a value. An alternative option was considered: a “sliding scale” where different fractions of a point would be awarded to a response depending on how close it was to the concern (a response within 2 turns would get .9 points, for example, while one addressed 6 turns later would get .5 points). While the capabilities of Sequence Viewer made it possible to impose this type of structure, in reality it would not be accurate: sometimes an answerer’s concern goes on for

several turns, with the interviewer offering only backchannels (such as “mm hmm” or “I see” in intervening turns); in the example below, the interviewer addresses the concern (“What’s the study about?”) as soon as she can without interrupting the answerer, but in the scheme discussed above, she would only get “credit” for addressing the concern within three turns.

Answerer: Ok well what—what's the study about?

Interviewer: [breath] Well,

Answerer: I have a thirteen- and a sixteen-year-old.

Interviewer: Uh huh,

Answerer: So what is it that I can help you with? [laughs]

Interviewer: Oh [breath], well, this is, um, just on how people make—
make medical decisions [breath].

3.3 Results

3.3.1 Prevalence of Answerer Concerns

This section describes the range of concerns expressed by answerers, with particular attention paid to differences across contacts with different outcomes. While concerns by answerers are common in survey introductions, they range in content and severity. Concerns are expressed with different frequencies in contacts with different outcomes.

Figure 3.3 shows the distribution of answerer moves in the dataset. A total of 7,342 answerer moves were coded (an average of 11.73 moves per contact, with contacts ranging from 4 to 245 moves). Over half (58 percent) of these contained no opportunity for interviewer tailoring (including answerer backchannels, such

as “uh huh,” greetings, and responses to standard interviewer questions, such as “This is a home” or “There are two adults in this household”).

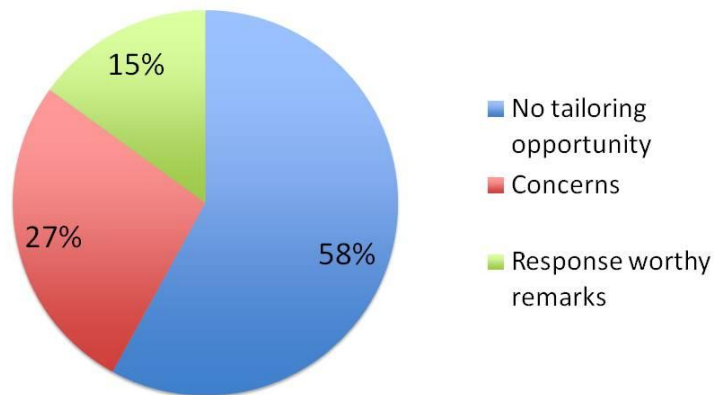


Figure 3.2: All Answerer Utterances

A substantial proportion (42 percent) of answerer moves provided an opportunity for interviewer responsiveness: 15 percent were conversation starters (described in Section 3.2), and about a quarter (27 percent) were concerns. This pattern varied depending on the outcome of the contact. While similar proportions of answerer moves were conversation starters in all outcomes (agrees 16 percent, refusals 14 percent, scheduled callbacks 16 percent), the proportion of moves which were concerns varied widely: 43 percent of answerer moves in refusal contacts were concerns, compared to 26 percent in scheduled callbacks and just 10 percent in agrees.

Overall, in 85 percent of the 626 coded contacts, answerers expressed some type of concern about either participating in the study, characteristics of the call (such as the volume), or the interviewer’s speech (such as the speed).

The likelihood that an answerer will express any concerns varies across contacts that result in agree, refusal, or scheduled callback. As shown in Table 3.2, close to half of those who ultimately agree express no concerns. In contrast, when contacts result in scheduled callbacks, the expression of some type of concern is nearly universal. Concerns are expressed in nearly all refusals as well.

Table 3.2: Proportion Contacts with One or More Concerns Expressed by Outcome

	Proportion contacts with one or more concerns expressed	N
Agree	.55	157
Refuse	.89	194
Scheduled callback	.99	275

Table 3.3 shows that the mean rate of concerns (number of concerns/number of answerer speaking turns) is highest for refusals (.35 concerns per turn) and lowest for agrees (.08 concerns per turn), while scheduled callbacks fall in the middle (.21 concerns per turn). The rates are significantly different between agrees and scheduled callbacks ($t[430]=-10.41, p<.001$) and between scheduled callbacks and refusals ($t[467]=8.57, p<.001$). These differences in rates occur despite the fact that refusal contacts are less than half the length of agrees and scheduled callbacks. This may be an indicator that refusers tend to express concerns, but are not swayed by the interviewer's response, or they express concerns that are difficult for the interviewer to address (ease of addressing different concerns will be discussed further in Section 3.3.5).

Table 3.3: Concern Rates by Outcome

	Mean concerns expressed (se)	Mean answerer turns (se)	Mean rate of concerns (se)
Agree	1.09 (.10)	13.03 (.61)	.08 (.01)
Refuse	1.88 (.10)	6.13 (.37)	.35 (.02)
Scheduled callback	2.16 (.10)	13.75 (.78)	.21 (.01)

The lower rates of concerns among agreeers begs the question as to why those who agree are so much less concerned than refusers or deferrers. I explored the possibility that their concerns have already been addressed in prior calls. Importantly, the only data available concerned call, not contact, number. While all human-to-human contacts are counted as calls, calls can also encompass messages left on answering machines, or calls where the household phone rings but no one answers.

In contacts where the outcome is “agree,” the number of calls does not have an impact on the number of concerns expressed. Answerers express 1.08 concerns if the contact is the first, second, or third call to their household, and 1.11 concerns if it is at least the fourth call, as shown in Table 3.4. This finding offers evidence that those who agree are not doing so because their concerns have been addressed on prior calls; they express the same number of concerns in later calls as earlier ones.

Table 3.4: Concerns Expressed in Agree Contacts, by Call Number

Call number	Mean number of concerns expressed	N (number of contacts)
1-3	1.08	77
4 or greater	1.11	80

Further evidence indicating that call number does not affect the number of answerer concerns comes from a comparison between agrees and scheduled

callbacks. Table 3.5 shows the distribution of call numbers among contacts in this dataset, and the mean call number, for each outcome, and demonstrates that the mean call number for agrees is not significantly different from that for scheduled callbacks ($t[430]=.89, p=0.37$). This could indicate that some answerers are simply more inclined to agree and less inclined to express concerns, an idea which will be discussed further in section 3.3.4.

Table 3.5: Distribution of Call Number by Outcome

Call number	Agree	Refuse	Scheduled callback	Total contacts
1	17% (27)	7% (14)	23% (63)	17% (104)
2	19% (30)	14% (28)	15% (41)	16% (99)
3	13% (20)	10% (19)	9% (25)	10% (64)
4	7% (11)	12% (23)	8% (22)	9% (56)
5	6% (9)	9% (17)	7% (19)	7% (46)
6 or higher	38% (60)	48% (93)	38% (105)	41% (258)
Mean call number	6.88	8.76	6.25	7.19
Total number of contacts in dataset	157	194	275	626

In some cases, scheduled callbacks result from an answerer’s willingness to say anything, including agreeing to a future call, in order to get off the phone. One motivator for these “false deferrals” may be politeness. As Brown and Levinson (1987) define it, politeness includes allowing one’s conversational partner to maintain positive face or “the want of every [person] that his wants be desirable to at least some others” (p. 322). By allowing interviewers to schedule a callback they have no intention of honoring, these answerers are being “polite” and pretending that they share the interviewer’s “want.” The role of politeness will be discussed further in Section 4.4.7.

Just as each contact to a household is assigned an outcome (agree, refuse, or scheduled callback), the ultimate outcome for each “case,” or household, is assigned. Analysis at the case level shows that for contacts in the corpus which are scheduled callbacks, approximately half (56 percent) are nested within cases where the household ultimately refuses to participate in the survey, demonstrating the polite “false deferral” phenomenon explained above. At the same time, close to half (42%) of contacts are in cases where the final outcome is participation. This supports the position that interviewers should schedule callbacks whenever possible, and that scheduled callbacks can often be seen as “salvaged” refusals.

3.3.2 Classification of Answerer Concerns

As shown in Table 3.6, the most common concern expressed across all contacts, regardless of outcome, was that interviewers were calling at an inconvenient time. Other common concerns included statements of general disinterest, queries about the length or purpose of the survey, issues related to the pre-notification letter (such as not receiving or not understanding it), and issues of comprehension (such as not hearing or understanding the interviewer).

The prevalence of different concerns varied across contacts that resulted in different outcomes. In contacts where the answerer ultimately agreed to participate, the most common concern expressed was one which was easily addressed by interviewers: questions about the purpose or content of the survey. Among the 272 contacts containing concerns that resulted in a deferral or scheduled callback, 73 percent contained a concern about the timing of the call—

also an issue which interviewers can (and did, as will be shown in Table 3.10) easily address, by offering to call again later. Indeed, addressing this concern in this way is what defines a scheduled callback. In contrast, in contacts that resulted in refusal, only 28 percent contained concerns about the call’s timing, while far more (55 percent) contained an expression of disinterest—a concern which is rarely addressed directly, as will be seen in Table 3.11. Refusers were also much more likely than deferrers to express concerns about being “taken off the list” (21 percent versus 1 percent, respectively).

Table 3.6 describes the percentage of concern-containing contacts, by outcome, that contain each specific concern. Eighty-five percent of all contacts (531 out of 626) had at least one concern expressed; column 1 shows the percentage of all concern-containing contacts that contained each concern, while columns 2–4 show the percentage of concern-containing contacts of each outcome (agree, refuse, scheduled callback) that contained each concern. Note that the percentages for individual concerns add up to more than 100 percent, since most contacts with concerns contained multiple concerns. An average of 2.1 concerns per contact (excluding contacts with no concerns) were expressed, and 58 percent of contacts with any concerns contained more than one concern.

Table 3.6: Prevalence of Concerns by Outcome in Contacts Containing any Concerns

Concern	% all Outcomes(n=531)	% agrees (n=86)	% refusals (n=173)	% SCBs (n=272)
Bad time	39	5	28	73
Not interested	22	2	55	9
Other person not available	19	3	11	35
Length of survey	16	25	5	19

Purpose/ content of survey	15	15	13	15
Harassment	7	0	19	2
Pre-notification letter	13	13	8	16
Comprehension (includes hearing/ speed)	10	14	6	12
Do not call list/ take me off list	7	1	21	1
Selection processes	6	4	4	8
Not suitable respondent	4	4	7	3
Other	4	0	6	4
Privacy/ confidentiality	4	4	3	4
Incentive	3	6	<1	2
% OF CONTACTS WITH ANY CONCERN	85	55	89	99

3.3.3 Conversation Starters

Beyond the expression of concerns, answerers often give opportunities for interviewers to tailor or personalize their conversation or deviate from a standardized script. I call these “conversation starters.” As Figure 3.3 showed, 15 percent of answerer moves contain conversation starters; while they are less prevalent than concerns, at least one such conversation starter appears in 54 percent of contacts.

As described in Section 3.2, examples of conversation starters include questions addressed to the interviewer, conversation peripheral to the task at hand, and any chance for an interviewer to acknowledge an answerer’s remark. Two examples of effective interviewer responses to conversation starters are below. In the first, the interviewer’s response to the answerer’s question show that she is contributing to discourse, or adhering to conversational norms by answering a question addressed to her; her second turn, “Oh, thank you,” is an expression of politeness. In the second, the interviewer’s response is more

substantial and demonstrates attention, adaptation to the answerer's comment, and quick thinking to build rapport with the answerer:

1) Answerer: Um, this is your job right?

Interviewer: Yes sir it is.

Answerer: Oh [laughs] oh, ok well we want you to keep your job.

Interviewer: [laughs] Oh, thank you.

2) Answerer: Just so you know, the next time Nebraska plays Michigan [laughs] we're going to root for Nebraska even though you're giving us fifty bucks.

Interviewer: That's all right. I'll root for Nebraska if you do the interview.

How's that?

Conversation starters can also include less overt comments by answerers, as demonstrated in the following two examples. In both of the examples below, the answerer responds to a nonstandardized question asked by the interviewer. However, interviewer responsiveness is different. In the first example, the interviewer acknowledges the answerer's response to her question, showing that she has heard the answerer. In the second example, the interviewer does not "take the bait" when given an opportunity to tailor her speech to the answerer's response; she does not acknowledge the answerer's "yes" but instead moves right into her introduction:

1) Interviewer: You may have received a letter regarding the study.

Answerer: Uh-huh.

Interviewer: You received a letter?

Answerer: Uh, yeah.

Interviewer: Oh, good.

2) Interviewer: We were wondering if you received our letter in the mail?

Answerer: Yes.

Interviewer: Well we're hoping, um, to have a few minutes of your time at your convenience.

The distribution of conversation starters is very different from the distribution of concerns. Recall from Section 3.3.1 that nearly all scheduled callbacks and refusals contained at least one concern, while only 55 percent of agrees did. In contrast, conversation starters are much more common in agrees. Interviewers should be aware that while concerns indicate wariness about participation, an answerer who offers conversation starters seems to be indicating engagement or interest and may be headed for agreement. Further discussion of these “green lights” can be found in Section 4.3.

Table 3.7: Proportion Contacts with Conversation Starters by Outcome

	Proportion contacts with conversation starters	N
Agree	.73	157
Refuse	.40	194
Scheduled callback	.53	275

Table 3.8 shows that, while the raw number of conversation starters in refusals is low, the mean rate (computed as number of conversation starters / number of answerer speaking turns and presented as column three in the table) is comparable to that in scheduled callbacks. Agrees stand out as having the highest rate of conversation starters compared to refusals ($t [349]=1.92, p=0.03$) and scheduled callbacks ($t [430]=3.78, p<0.001$).

Table 3.8: Conversation Starter Rates by Outcome

	Mean (se) conversation starters	Mean (se) answerer turns	Mean rate of conversation starters (se)
Agree	1.59 (.14)	13.03 (.61)	.12 (.01)
Refuse	.60 (.07)	6.13 (.37)	.09 (.01)
Scheduled callback	1.51 (.18)	13.75 (.78)	.08 (.01)

3.3.4 Interviewer Responsiveness Scoring

As discussed in Section 3.2, an interviewer responsiveness score for each contact is computed as follows:

$$\frac{[\text{Concerns addressed immediately} + (\text{Concerns addressed later}/2) - \text{Concerns never addressed} + \text{Conversation starters addressed immediately} + (\text{Conversation starters addressed later}/2) - \text{Conversation starters never addressed}]}{(\text{Total concerns expressed} + \text{Total conversation starters})}$$

Thirty-six contacts with no conversation starters or concerns were dropped from this analysis. In addition, concerns are not counted in this analysis if they are expressed as the last turn of a contact—that is, if the answerer hangs up immediately after stating the concern, thus leaving the interviewer no chance to respond.

Groves and McGonagle (2001) found that interviewers who were trained in addressing specific concerns were more successful than their counterparts who did not receive this training. Along these lines, and as discussed in Section 3.1, I hypothesize that interviewers who are responsive to answerer concerns and conversation starters will be more successful in individual invitations than interviewers who are not. This is supported: responsiveness scores are higher for agrees than refusals ($t[315], =12.65 p<0.001$). The negative mean

responsiveness score for refusals indicates that interviewers are not responding to concerns or conversation starters in these contacts, either because they are not performing well or because the concerns presented are difficult to respond to (such as “I’m not interested.”). However, the highest responsiveness scores are seen in scheduled callbacks, as shown in Table 3.9. The mean responsiveness score for scheduled callbacks, .80, is significantly higher than that for agrees, .63 ($t[406]=-3.02, p<0.01$). The differences in responsiveness scores by outcome reported here are significant at the $(.05/3)= 0.16$ level when a Bonferroni correction is applied.

Table 3.9: Mean Responsiveness Score by Outcome

	Mean (se) responsiveness score
Agree	.63 (.05)
Refuse	-.32 (.05)
Scheduled callback	.80 (.03)

The pattern shown above in Table 3.9 indicates that responsiveness does not necessarily convert answerers to agreement. As shown in Tables 3.3 and 3.5, answerers to agree contacts are less likely than those to scheduled callbacks to express concerns, despite having received the same number, on average, of prior calls. Therefore, it seems plausible that answerers who ultimately agree are simply more inclined to do so, and that they express fewer concerns for some reason other than the fact that their concerns have already been addressed. The reason may be one (or several) of the four categories of reasons for nonresponse outlined by Groves et al. (2009): the social environment, the person, the interviewer, or features of the survey design.

Although clear cause and effect patterns cannot be established in this study (could it be that answerers to scheduled callbacks simply more inclined to defer, and agreeers more inclined to participate, regardless of the interviewer’s behavior?), it appears that addressing concerns effectively can be a powerful technique in converting potential refusals to scheduled callbacks—which ultimately become agrees in future calls close to half of the time, as mentioned in Section 3.3.1.

Interviewer responsiveness to concerns and conversation starters follows the same pattern across outcomes. Interviewers are most responsive (to both concerns and conversation starters) in scheduled callbacks, and least responsive in refusals. Table 3.10 shows responsiveness scores decomposed into an “addressing concerns” portion, computed as $(CA1+(CAL/2)-CAN)/\text{total concerns}$ expressed, and a “responding to conversation starters” portion, calculated as $(CSA1+(CSAL/2)-CSAN)/\text{total conversation starters}$.

While responding to conversation starters may be less critical than addressing concerns, an interviewer’s overall responsiveness (to both concerns and conversation starters) is important to the success of a contact.

Table 3.10: Decomposing Responsiveness Scores

	Mean (se) Addressing Concerns score	Mean (se) Responding to Conversation Starters score
Agree	.76 (.07)	.54 (.07)
Refuse	-.37 (.06)	.10 (.11)
Scheduled callback	.86 (.03)	.64 (.06)

It should also be noted that responsiveness is a contact-level variable, which cannot be used to draw conclusions about interviewers in this dataset. Of the 58

interviewers represented in this dataset, 32 had at least 10 contacts. For these 32 interviewers, a mean responsiveness score was computed for all of their contacts; these means ranged from -.18 to .72. The standard deviations for individual interviewers ranged from .08 to 1.59, with an average value of .74. This indicates that often, responsiveness scores are not consistent within interviewers across contacts. It can be concluded that interviewer responsiveness, then, is not entirely in the hands of the interviewer; it also depends in part on the speech and the actions of the answerer. Whether an interviewer is responsive to a concern depends largely on the type of concern presented by the answerer, as will be discussed in Section 3.3.5; difficult to address concerns such as “I’m not interested” or “I don’t do surveys” are most prevalent in refusals and are addressed infrequently.

3.3.5 Addressing Specific Concerns

This section explores interviewers’ responses to specific concerns, with particular attention paid to answerers’ statements of disinterest. As shown above in Table 3.6, the distribution of concerns expressed varies widely by ultimate contact outcome, with concerns about the timing of the call dominating scheduled callbacks. Questions regarding the length of the survey are the most frequently expressed concern among contacts that result in agreement—paralleling the finding by Schaeffer, Garbarski, Maynard and Freese (2011) that when answerers ask about interview length, acceptance is more likely—while expressions of disinterest are the most common concern in refusals.

As Table 3.11 shows, concerns are also not only expressed, but are also addressed, at different rates. For example, while 93 percent of concerns about the survey’s length were addressed, only 51 percent of concerns about prenotification letters were. Additionally, the same concerns are addressed at different rates across different outcomes. The lower rates of address in refusals compared to other outcomes (for example, concerns about the Do Not Call list, pre-notification letters, or not being a suitable respondent) point to the importance of interviewer responsiveness to answerer concerns.

Table 3.11: Concerns Addressed by Outcome

Concern	Total addressed (as % of # expressed)	% addressed in agrees	% addressed in refusals	% addressed in SCBs
Bad time	88	75	53	94
Not interested	6	0	5	12
Other person not available	85	40	63	92
Length of survey	93	97	90	90
Purpose/ content	93	96	88	95
Pre-notification letter	51	60	31	55
Comprehension	82	86	82	78
Do not call list/ take me off list	62	100	35	50
Harassment	5	0	5	0
Not suitable respondent	52	67	31	75
Selection process	71	100	25	95
Privacy/ confidentiality	91	100	80	91
Incentive	94	100	100	83
Other	41	0	25	60

The different responsiveness scores across different outcomes may be driven by the ease with which an interviewer can respond to different concerns.

For example, concerns about call timing or survey length (most common in agrees and scheduled callbacks, respectively) are relatively straightforward for interviewers to address:

- 1) Answerer: But you're calling at a bad time because we have company.
Interviewer: Oh, oh, I see. Is there a better time that we could call back?
- 2) Answerer: How many minutes is the survey?
Interviewer: It takes maybe ten minutes. It's pretty short.

Sweeping statements of disinterest, such as “I’m not interested” or “I don’t want to participate,” and personal policies such as “I don’t do surveys,” are much harder for interviewers to respond to. Because of the small number of personal policy concerns, all concerns of this nature were coded as “expression of disinterest.” Such concerns were expressed in 135 contacts and were most common in refusals, as shown in Table 3.12.

Table 3.12: Distribution of Contacts Containing Statements of Disinterest by Outcome

Outcome	% of contacts with this outcome containing statement of disinterest	% of all contacts containing statement of disinterest that have this outcome
Agree	2%	2%
Refuse	55%	79%
Scheduled callback	9%	19%

As shown in Table 3.11, “not interested” concerns are among the least frequently addressed: only 6 percent of instances of these concerns are addressed by interviewers. Often, this is because interviewers do not have a chance to address the concerns: in 26 percent of contacts containing an expression of disinterest, the answerer did not say anything after the statement

of disinterest before hanging up, showing no willingness to react to an interviewer's response. These statements, then, should be interpreted by interviewers as "red flags" that indicate that unless drastic action is taken, the contact is about to be terminated.

Statements of disinterest are often presented in combination with another, more easily addressable concern, and interviewers often default to addressing these concerns, rather than the answerer's lack of interest. In the following examples, the interviewer chooses to respond to other issues brought up by the answerers (repeated calls in the first example and a misunderstanding of the purpose of the call in the second):

1) Answerer: I really don't want to take it. So I need you to take me off the list or quit calling here because I don't have time to do a survey. I've already declined. They've called me like three or four times. I told them the last time that I just wasn't interested in doing it.

Interviewer: Oh, I do apologize ma'am for all the calls. We are actually coming to the end of our study and we really do need representation from your area.

2) Answerer: Yeah well I won't be interested in that. I don't even know what it's about. And then plus I'm tired of telemarketers calling here.

Interviewer: Oh I completely understand ma'am. You know a lot of times we do get confused with telemarketers. We are not telemarketers.

In 21 percent of contacts containing an expression of disinterest, interviewers seek out an addressable concern by asking why the answerer does not want to

participate. Interviewers may do this as a means to establish common ground or a mutual understanding with the answerer, but Pondman (1998) found that this type of query prompts answerers to verbalize or repeat their reasons for not wanting to participate and rarely leads to conversions; interviewers at the University of Michigan (where all studies in this dataset were fielded) are advised to “break the habit of asking what the concerns are” (Joe Matuzak, personal communication, 2011). However, this is not an uncommon technique among interviewers. Such an approach can be beneficial, as it sometimes prompts the answerer to express a concern that the interviewer can easily address. Indeed, among the 28 contacts where interviewers responded to a statement of disinterest by asking why the respondent did not want to participate, interviewers managed to turn six (21 percent) of them into scheduled callbacks and one (3 percent) into an agree, often by following their question with a barrage of information intended to assuage myriad possible concerns, as in this example:

Answerer: Uh, you know, I’m not interested in that.

Interviewer: Ok, are there any concerns you have? This study is one of the most important studies in the country. It’s looked at by the Federal Reserve Board. Your number was chosen to represent your part of California and you really can’t be replaced in the study. It’s just general opinion questions.

Answerer: Ok, all right, let’s do it.

Still, 76 percent of contacts where interviewers responded to a statement of disinterest by asking answerers to elaborate on their concerns resulted in refusal.

This shows not much improvement over the refusal rate of 79 percent in any contacts where an expression of disinterest was expressed (whether interviewers responded or not).

Instead of probing for a specific concern, other techniques for responding to expressions of disinterest included ignoring, acknowledging, or rejecting them, as in the examples below. It cannot be said with certainty that these responses improve response likelihood; instead, they are presented as options for interviewers to consider in lieu of asking disinterested answerers to elaborate on their concerns.

Here the interviewer bypasses the answerer's statement of disinterest and moves into the household listing:

Answerer: I don't think I'm interested.

Interviewer: We can just do the first part to determine who is eligible and then after that we can just set up an appointment to call back later. It just takes two minutes to find out who in your household the computer will pick to participate. So what's your first name?

In the examples below, rather than ignoring the statement of disinterest, the interviewer acknowledges it directly. Faber and Mazlish (1980) suggest that acknowledging and naming negative feelings when they are expressed, rather than rejecting or downplaying them, is an effective tactic for engaging children in distress; it appears that some interviewers employ this technique with potential respondents as well:

1) Answerer: Well, I ain't interested.

Interviewer: I know you're saying you're not interested, but I'd be more than willing to talk about the study with you right now so you can familiarize yourself with it.

2) Answerer: I just don't want to do it is what it comes down to.

Interviewer: Yeah, I understand that part. Right.

Sometimes interviewers offer an explanation as to why the answerer's policy of nonresponse does not apply in this situation, as in this example:

Answerer: I just don't like to participate in phone surveys.

Interviewer: Well this is actually not a typical phone survey.

Addressing the lack of interest—that is, treating a statement of disinterest or a nonresponse policy as a legitimate and addressable concern, rather than asking answerers to elaborate on the reasons for their disinterest—may help interviewers to avoid the phenomenon observed in this dataset and discussed by Pondman (1998), where answerers who are probed to express their reasons for not wanting to participate are less likely to be converted to agreement. Other examples of rebuttals to statements of disinterest included:

1) Interviewer: Most people find this a pretty interesting study and this is really the first time it's ever been done.

2) Interviewer: A lot of people who haven't wanted to do it did participate in it and found it quite interesting.

3.3.6 Predictions of Contact Length and Outcome using Responsiveness Score and First Impressions

Recall that the key finding from Chapter 2 was that contacts where interviewers are rated as less scripted are more likely to result in success. In this section, I test the hypothesis that contacts where interviewers have both lower ratings of scriptedness and higher responsiveness scores will be the most likely to result in agreement. Data from the listeners' study discussed in Chapter 2 are analyzed in conjunction with interviewers' responsiveness scores.

Analyses were conducted on a subset of 283 of the 626 coded contacts. All contacts in this subset were included in the listeners' survey described in Chapter 2, where excerpts from the first interviewer turn were rated on twelve personality characteristics by respondents to a Web survey. All contacts in this subset either resulted in agreement (n=118) or refusal (n=165).

In two separate logistic regression models, each using a single predictor (both of which accounted for clustering by interviewers using multilevel, mixed effects modeling), both ratings of scriptedness ($z = -2.59$, $p = 0.010$) and an interviewer's responsiveness score ($z = 8.01$, $p = 0.000$) were related to the likelihood of the contact resulting in agreement. However, when a multiple logistic regression model was constructed (controlling for interviewer experience and exposure length in the listeners' study, and accounting for clustering by interviewer) using both scriptedness and responsiveness as predictors of outcome, only responsiveness was associated with the likelihood of agree, indicating that

responsiveness across the entire contact is more important to outcome than scriptedness in the initial turn. Results from this model are shown in Table 3.13.

Table 3.13: Model Predicting Log Odds of Agree

	Coefficient	SE	Z	P
Scripted	-.31	.50	-0.62	0.537
Responsive	1.94	.25	7.85	0.000
Experience	.0001	.0001	1.06	.291
Exposure length	.007	.02	.29	.768

Contact length of the introduction (the number of seconds from the first “hello” until either the hang-up or the first question of the interview) was also used as a measure of a successful contact. Groves and Couper (1998) observe that interviewers who are able to prolong interactions will be more likely to have success, and it was observed in this dataset that the mean length of agree contacts is 170.69 seconds, significantly longer than that for refusals (48.47 seconds, $t[281]=17.25$, $p<.001$). A model predicting contact length was constructed, with scriptedness and responsiveness as predictors, controlling for exposure length (the length of the rated excerpt) and interviewer experience (which is associated not only with contact length, but also likelihood of agreement) and accounting for clustering by interviewer. Results shown in Table 3.14 mirrored those from the logistic model predicting the likelihood of agreement described above. An interviewer’s responsiveness is associated with a longer contact length, while an initial rating of scriptedness is not.

Table 3.14: Model Predicting Contact Length

	Coefficient	SE	Z	P
Scripted	54.28	143.78	0.38	0.706
Responsive	597.01	54.08	11.04	0.000
Experience	.026	.04	0.74	.459
Exposure length	11.36	6.17	1.84	0.240

When an interaction of scriptedness and responsiveness is added to these models, it is not significant.

There seems to be a “tipping point” above which ratings of scriptedness have an impact on not only agreement, but the likelihood that an interviewer will be responsive. Scriptedness was rated in the listeners’ study on a six-point scale, with a median rating of 3.78. At a point slightly below the median, 3.7, we observe substantial shifts in the impact of scriptedness. Thirty-seven percent of contacts with scriptedness scores at or above 3.7 result in agreement, compared to 50 percent of contacts with scriptedness scores below 3.7 ($t [281] = 2.2$, $p = 0.03$). The difference in agree rates persists when contacts with no tailoring opportunities are removed from the analysis (high scripted: 35 percent agree, low scripted: 47 percent agree; $t [253] = 2.0$, $p = 0.02$). When we perform this comparison by splitting contacts at the median of 3.78, 46 percent of low scripted and 37 percent of highly scripted contacts result in agreement, a difference which is not significant ($t [253] = .79$, $p = 0.22$).

Similarly, comparing contacts with scriptedness ratings above and below 3.7 shows a significantly higher proportion of contacts with high responsiveness scores (above the median), compared to low responsiveness scores (at or below the median). Eliminating contacts with no tailoring opportunities, 57 percent of low scripted contacts have high responsiveness scores, compared to 46 percent of highly scripted contacts ($t [253] = 1.6$, $p = 0.05$). In comparison, when scriptedness is split at the median, 52 percent of low scripted and 48 percent of high scripted contacts have high responsiveness scores ($t [253] = .69$, $p = 0.25$).

This indicates that some interviewers may make use of two techniques (low scriptedness and high responsiveness) to improve cooperation, while other interviewers used neither.

The agree rates in Table 3.15 demonstrate that the optimal combination of behaviors for interviewers is low scriptedness and high responsiveness. At the same time, it is preferable to be highly scripted and highly responsive than less scripted and less responsive: among the 55 contacts where interviewers had a mean scriptedness rating above 3.7 and a responsiveness score at or above the median score of .083, 62 percent resulted in agreement. In comparison, among the 85 contacts with high scriptedness ratings (at or above 3.7) and low responsiveness scores (below .083), only 9 percent resulted in agreement.

Table 3.15: Agree Rates by Scriptedness/Responsiveness Quadrant

		Responsiveness	
		High	Low
Scriptedness	High	64%	9%
	Low	73%	14%

Interviewers who are less scripted and more responsive are marginally more successful than those who are more scripted and more responsive ($t[154] = .17$, $p = 0.09$). The ideal combination, clearly, is being unscripted and responsive, but even highly scripted interviewers can redeem contacts by being responsive.

However, the effect of being highly responsive, regardless of level of scriptedness, is strong: overall, 68 percent of contacts with highly responsive interviewers resulted in agreement, while 11 percent of contacts where interviewers received low responsiveness scores did.

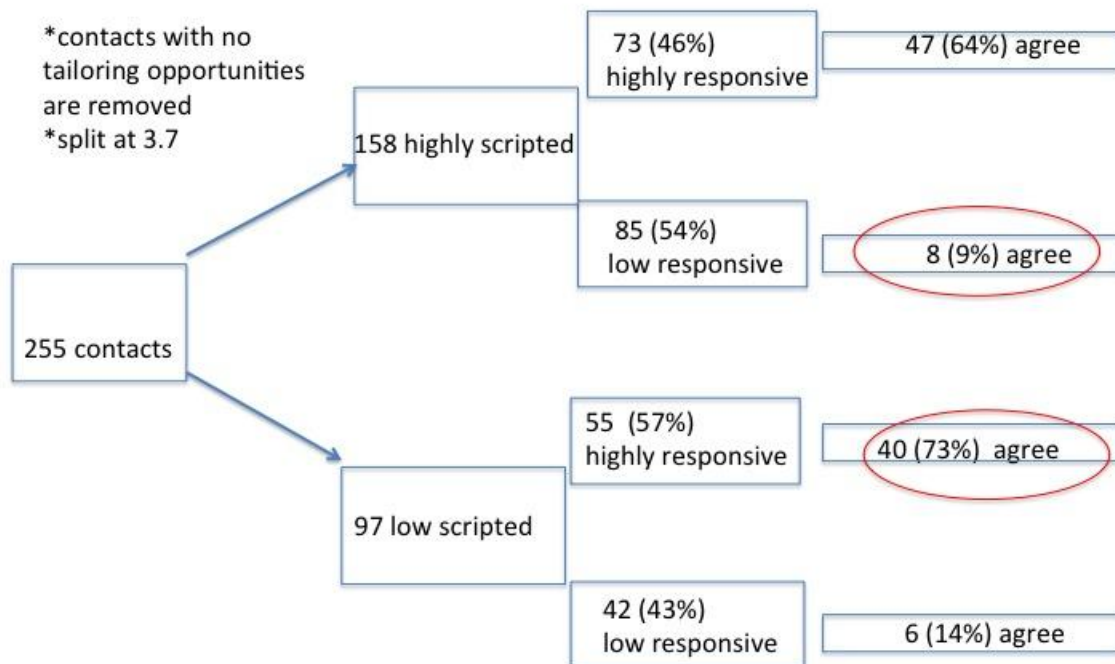


Figure 3.4: Scriptedness/ Responsiveness Paths

3.3.7 Comparison to Practitioners' Study

The importance of responsiveness to an interviewer's success was clearly understood by the 44 survey practitioners surveyed in the study discussed in Chapter 2. "The ability to address concerns expressed by potential respondents" was considered "extremely important" to an interviewer's success by 83 percent of practitioners surveyed, behind only "the initial impression an interviewer gives to sample members (88 percent)" and "professional demeanor when talking to potential respondents" (86 percent). However, when it comes to training, practitioners seem to underemphasize the importance of addressing concerns relative to other interviewer skills, mostly related to question administration. Only

78 percent of practitioners surveyed reported that “responding appropriately to arguments or concerns expressed by potential respondents” was a primary training focus in their organization, in comparison to 98 percent of respondents who named “administering interviews” and “importance of data quality” as primary foci, and 93 percent who named “standardized interviewing skills.”

Along similar lines, practitioners who had responsibility for interviewer hiring were asked to rate how important “beliefs about how the individual will interact with respondents and potential respondents” were to their hiring decision. While not ranked as highly as “hard” job skills such as English fluency, literacy level, and speaking voice, nor as highly as attributes such as reliability and trainability, these perceptions were “extremely important” to 58 percent of practitioners surveyed. This ranked above a candidate’s performance in training exercises, experience, references from previous employers, and beliefs about how the candidate would get along with other team members.

3.4 Conclusions

This study clearly demonstrates the importance of responsiveness on the part of a telephone interviewer. Responsive behavior can overcome the negative effect of first impressions of an interviewer as scripted; even interviewers who are initially scripted, if they can keep the answerer on the phone, will be more successful if they are responsive to answerer concerns and conversation starters.

The work of an interviewer in being responsive varies greatly. Answerers who ultimately agree tend to express fewer concerns, but make more conversation

starters. In contacts that result in refusal, answerers express a relatively high number of concerns in a shorter time period, and often do not give the interviewer a chance to respond. Answerers who defer have a similar rate of concerns to refusers, but they are spread out over a longer contact. Interviewer responsiveness to concerns plays the greatest role in attaining scheduled callbacks.

The types of tailoring opportunities (conversation starters and concerns) expressed by answerers can offer the interviewer a clue as to where the contact is headed. Concerns about the purpose or content of the survey, or the length of the interview, are most common in contacts where the answerer ultimately agrees to participate, as are conversation starters. If an answerer states that the interviewer has called at a “bad time,” the contact can often be converted into a scheduled callback, although some wariness about the true intentions of the deferring answerer is recommended, since over half (56 percent) of households where a callback was scheduled ultimately refused. Expressions of disinterest or “personal policies,” such as “I don’t do surveys,” should be viewed as red flags and handled with caution. Interviewer responses to these concerns are often ignored and followed or interrupted by hang-ups. In responding to this type of move, probing for more information or asking the answerer to elaborate on his or her concerns may not be an interviewer’s best strategy, as it prompts answerers to descend into a spiral of negativity. Interviewers may have a better chance of success if they treat and address “I’m not interested” as a concern in and of itself, rather than view it as a symptom of another concern.

3.5 Limitations

A limit of this study is the lack of clear cause and effect patterns. As discussed in section 3.3.4, it is impossible to know in an observational study such as this one if answerers who defer are simply less prone to outright refusal regardless of interviewers' responsiveness, or if it is interviewers' responsive behavior that converts these answerers to scheduled callbacks instead of refusals.

Another limitation of this work is that it may not be applicable to all surveys. The contacts in this corpus were selected from five studies conducted at the University of Michigan and were not intended to be representative of all telephone surveys.

At the same time, the richness of the qualitative data analyzed here should be further explored as likely having some relevance for other surveys. While the content of concerns may differ slightly among, for instance, sample members from list samples, previous respondents (or nonrespondents) to panel surveys, or potential respondents to establishment surveys, I suspect that the importance of an interviewer's responsiveness will not change. Future work replicating this type of methodology on a larger dataset, encompassing contacts from different survey institutions and types of surveys (such as the examples listed above), could help to solidify these findings.

Additionally, the analysis including both first impressions and responsiveness was limited by the data collected in the listeners' study. While only agrees and refusals were analyzed in that study by design (to facilitate cleaner comparisons),

future work looking at first impressions and responsiveness in a corpus which includes scheduled callbacks is recommended.

It is also important to point out the heterogeneous nature of contacts classified as “scheduled callbacks.” Some of these contacts are genuine deferrals, where the answerer honestly does not have time and initiates the callback. Others reflect an interviewer’s last-ditch effort to save the case, blurting “We’ll call you back!” even as the answerer hangs up. Analyses of responsiveness conducted on a more fine-grained division of scheduled callbacks could be very informative. For example, is ultimate agreement more likely when an answerer initiates a scheduled callback rather than waiting for the interviewer to do so, or when an answerer expresses a concern in the form of a question rather than a statement?

Finally, future work might also look further at the different impact of responding to concerns versus conversation starters. Even more granular analysis could look at the impact of addressing versus ignoring particular concerns, or particular types of conversation starters.

Chapter 4: Conclusion

4.1 Introduction

This chapter will summarize the findings from the previous chapters, as well as discuss practical implications from these studies for survey practice, and make suggestions for future research on telephone interviewer speech and behaviors.

4.1.1 *Review of Methods*

This dissertation discussed three studies. The listeners' study was a Web survey conducted among over 3,000 members of a commercial online survey panel. Listeners heard brief excerpts of interviewer speech from five actual telephone survey introductions and rated twelve characteristics of the interviewer. For five other excerpts, listeners rated the likelihood that the contact resulted in agreement.

The second study, also an online survey, was conducted among research practitioners, all of whom are responsible for hiring and/or training telephone interviewers. This study explored practitioners' opinions about the attributes of a successful interviewer, as well as their priorities in hiring and training interviewers.

Finally, the tailoring study involved detailed coding of the transcripts of telephone survey introductions, capturing concerns and conversation starters

presented by answerers, as well as interviewer responses to both concerns and conversation starters.

4.2 Summary of Findings

This dissertation was based on a model which proposed that when, excluding “hard-core” nonrespondents, an initial voice which was competent and warm could get an interviewer “over the hump” of an interaction—that is, past an immediate refusal. According to this model, after giving a positive initial impression, an interviewer who was responsive to an answerer would have more success in persuading the answerer to comply with the survey request. I also hypothesized that survey practitioners would place emphases on the attributes of interviewers that were rated highly in successful contacts.

4.2.1 Findings on Distinction of Person Perception Dimensions

Support for my hypotheses was mixed. As discussed in Chapter 2, survey practitioners believed that characteristics along a “competence” dimension, including being competent, confident, and professional, were more important to interviewers’ success than “warmer” traits, such as being friendly and enthusiastic. This suggests that common wisdom among practitioners distinguishes between warmth and competence of interviewers. In contrast, and also discussed in Chapter 2, judges listening to brief excerpts from the start of survey introductions did not distinguish between the “warmth” and “competence” dimensions of person perception; instead, ratings of all nine positive characteristics rated by judges (competent, confident, knowledgeable, professional, friendly, genuine, enthusiastic, natural-sounding, and pleasant to

listen to) were highly correlated, indicating that interviewers who are positively perceived on one dimension are positively perceived on the other.

4.2.2 Associations Between First Impressions, Vocal Characteristics, Predicted Outcomes, and Actual Outcomes

For a separate set of five contacts—that is, not the contacts for which they provided ratings—judges also rated the likelihood that a contact resulted in success. While ratings of positive characteristics were predictive of other judges' ratings of the likelihood that contacts resulted in success, neither these likelihood ratings nor ratings of any positive characteristics were predictive of the actual outcome of the contacts (see Chapter 2). This demonstrates that the first impression of an interviewer may not be the most important determinant to a contact outcome, despite practitioners' belief that “the initial impression an interviewer gives to sample members” is of high importance to an interviewer's overall success, as are answerers' impressions of the interviewer as competent, confident, genuine, and knowledgeable.

As discussed in Chapter 2, ratings of interviewer characteristics were associated with vocal characteristics, with interviewers who spoke more quickly and at a lower pitch in the brief excerpts to which listeners were exposed being rated more highly on nearly all positive characteristics (the only exception being friendliness, which was associated with higher pitch). However, in contrast to Benkí, Broome, Conrad, Groves, and Kreuter's (2011) finding that speech rate and fundamental frequency over the course of an interviewer's introduction can affect response likelihood, this

study found that interviewer speech rate and pitch in the initial interviewer turn have no impact on a contact's success.

While vocal characteristics and ratings of positive personality characteristics were both unrelated to contact success, ratings of an interviewer's scriptedness were significantly negatively associated with agreement, indicating that an interviewer who comes across to the listener as less scripted in the initial seconds of a contact has a greater chance of success.

4.2.3 Contrast Between Practitioners' and Listeners' Studies

The negative association between scriptedness and success discussed in Chapter 2 contrasted with the dominant belief among survey practitioners that the ability to follow an introductory script is important to an interviewer's success. Similarly, practitioners placed limited emphasis on training interviewers to "ad lib" or deviate from scripts, but comparatively more importance on training interviewers to follow introductory scripts, despite literature finding that interviewers who are allowed to improvise their introductions have more success than those who are required to follow a verbatim script (Houtkoop-Steenstra and van den Bergh 2000; Morton-Williams 1993).

Further reflection leads to the conclusion that judges and practitioners may be perceiving the word "scripted" differently based on the context. In a brief excerpt, judges' ratings of scriptedness are likely based on whether the interviewer sounds robotic, uses unnatural intonation, or sounds like she is reading. For survey practitioners, a "scripted" introduction may well be one where the interviewer touches on all the key points from an introductory agenda: name,

affiliation, sponsor, purpose of the study, confidentiality assurance, and incentive. From this perspective, an interviewer could speak very naturally, and even have “personalized” the script by presenting these elements in her own voice, but by covering all of these topics, she is deemed to have “followed the script.” Suggestions for future research to resolve this possible conflict between raters and practitioners will be discussed in Section 4.4.

4.2.4 Interviewer Responsiveness

While an initial impression of an interviewer as non-scripted is important, findings from the tailoring study discussed in Chapter 3 found that an interviewer’s ability to be responsive to an answerer over the course of the survey introduction is far more important to her success than any initial impression.

Two components of interviewer responsiveness were explored: addressing answerer concerns, and responding to conversation starters presented by answerers. Answerer utterances were coded to indicate if they contained one of thirteen specific concerns, including “bad time,” “purpose,” or “not interested,” or provided the interviewer with a conversation starter, either by answering a question posed by the answerer or engaging in conversation peripheral to the task at hand. Similarly, codes were assigned to interviewer utterances to indicate if the interviewer appropriately addressed a concern (for example, by responding to “I’m too busy” with “We can call you back later”) or “took the bait” and engaged in conversation when a conversation starter was presented.

For each contact, the interviewer received a score reflecting if she responded to concerns and conversation starters immediately, after a delay, or not at all. Scores varied by contact outcome, with a negative mean responsiveness score among refusals and the highest mean responsiveness score in contacts where the interviewer scheduled a callback.

Analyses revealed that while overall responsiveness by an interviewer is important, an interviewer's ability to address concerns trumps her responses to conversation starters in persuading an answerer to participate.

4.2.5 Scriptedness and Responsiveness

Interviewers who start out as scripted but are highly responsive as the answerer raises concerns or presents conversation starters are nearly as successful as their counterparts who begin with a low level of scriptedness and then act responsively. Thus, while being less scripted can help interviewers get past the initial "hump," in contacts that survive the initial stage, interviewer responsiveness is crucial to success.

4.3 Recommendations for Telephone Interviewing Practice Based on these and Other Studies

The findings discussed above, coupled with other findings in the survey methodology literature, can be applied to the practice of telephone interviewing to improve response rates. This section will discuss applications of these results for survey practice.

Recommendation #1: Train interviewers to switch gears from conversational introductions to standardized interviews.

The finding (see Chapter 2) that judgments of an interviewer's scriptedness in the initial seconds of a contact are negatively associated with contact success should be considered by those responsible for hiring, training, and monitoring interviewers. Interviewers should be encouraged to make their speech as natural as possible, through the use of intonation patterns and word selection. Interviewers can be exposed to contacts with both high and low ratings of scriptedness to make clear the difference.

While interviewers may be required to mention particular points in their introduction or even to follow a verbatim introductory script, they should be trained to sound as conversational as possible, particularly at the start of their introduction. Both Houtkoop-Steenstra and van den Bergh (2000) and Morton-Williams (1993) found that interviewers who were allowed to adapt their introductory script had greater success.

Further, work by Conrad, Broome, Benkí, Groves, Kreuter and Vannette (2010), which uses the same corpus of introductions from which the introductions in the listeners' study were culled, demonstrates that a moderate use of fillers such as *um* and *uh* by interviewers can lead to greater success—possibly because these interviewers sound like they are engaged in a natural conversation, rather than following a script.

It could behoove survey organizations to conduct research around which elements of speech (increased use of fillers, maintenance of natural intonation

patterns, and “on the spot” adaptation of scripts) can be taught, and then to focus on training interviewers to use these techniques.

Beyond the introduction, the issue of standardized interviewing, and what departures from verbatim interview scripts can mean for data quality, is the subject of much debate. Schober and Conrad (1997) and Conrad and Schober (2000) found clear evidence that “conversational” interviewing, or less rigid adherence to interview scripts, can enhance data accuracy. Along similar lines, Dykema, Lepkowski, and Blixt (1997) found that interviewer deviations from standard question wording had minimal impact on respondents’ reports. Still, “reading the questions exactly as worded” is a tenet of interview administration which is upheld and enforced in most survey organizations, and it is clear from results of the practitioners’ study that standardized interviewing skills are a high priority in nearly all organizations. Because emphasizing the need to read questions in a standardized manner may seem in conflict with emphasis on less scripted delivery of introductions, interviewers need to be trained to “wear two hats.” It needs to be made explicit to interviewers that there are two distinct (but, arguably, equally important) elements of the phone component of their job, each requiring a different style of speech and interaction. In the introductory or persuasive portion, scriptedness may be a liability, and the ability to “think on one’s feet” to respond to answerers is an asset. In contrast, in the interviewing portion, deviating from a script may have ramifications for data quality, or at the very least, will represent a lack of adherence to the organization’s procedures. Interviewers should be trained to “switch gears” between these two speech

styles, and perhaps even be encouraged to acknowledge to respondents that their delivery of the questions will sound different from their introduction.

Recommendation #2: Train interviewer speech rates; consider implementing hiring criteria around vocal pitch.

It may be a worthwhile investment by research organizations to place greater emphasis on interviewers' vocal characteristics. Speech rate in particular is something that can be trained and monitored. Findings in Chapter 2 indicate that interviewers who speak faster in the initial excerpt are rated more highly on all positive characteristics. While speech rate in this initial excerpt was not associated with the success of a contact, attention should be paid to the work of Benkí et al. (2011) finding that a rate of 3.5 words per second during the introduction is ideal in obtaining cooperation. This rate is fast enough to sound self-assured, but not so fast as to be incomprehensible. Notably, this is higher than the rate of two words per second, which is often suggested anecdotally as an ideal speech rate for delivering interview questions. Just as interviewers may be able to be trained to speak in a less scripted manner during an introduction and more so during an interview, they can be trained to slow down their speech noticeably after the introduction. This type of training could be implemented with the use of software measuring speech rate and displaying it to the interviewer as she talks (similar to speed clocks which show drivers their current speed); interviewers could slow down or speed up their speech in response, and eventually learn how it feels and sounds to speak at a rate of 3.5 (for an introduction) or 2.0 (for question delivery) words per second.

Additionally, while interviewer fundamental frequency in this study was not related to contact outcome, it was related to perceptions of interviewers. On eight of nine positive characteristics, female interviewers with lower voices received higher ratings than their higher-voiced counterparts. Benkí et al. (2011) found lower pitch over the course of a contact to be associated with success. If further research can substantiate these findings, survey organizations may want to limit their hiring to interviewers who are capable of hitting an optimal pitch range (or avoiding a pitch range which is associated with negative impressions and lower success).

Recommendation #3: Emphasize responsiveness to answerer concerns.

While scriptedness and vocal characteristics in the initial seconds of a survey introduction are important (see Chapter 2), it is clear from my tailoring study (see Chapter 3) that an interviewer's ability to be responsive to answerers is absolutely critical. Practitioners agree that responding to sample members' concerns is an important ability of a successful interviewer, and most emphasize this in training, though not as much as other elements. Practicing by interviewers of appropriate responses to common answerer concerns should be a top priority in interviewer training; as Groves and McGonagle (2001) demonstrated, interviewers trained in effective responding have greater success.

Recommendation #4: Train interviewers to be aware of and respond to both "red flags" and "green lights" from answerers.

Some answerer comments should be viewed as "red flags," or warnings that the contact is about to end. Particular attention should be paid to concerns in the

“I’m not interested” category. Interviewers often treat statements in this category as symptoms of a different concern and, instead of addressing the answerer’s stated lack of interest, attempt to uncover a more addressable concern, such as a lack of time or worries about privacy. Findings by Pondman (1998), and results discussed in Section 3.3.5, show that asking answerers to elaborate on this type of remark (as in, “May I ask why you don’t want to participate?”) can have disastrous consequences, sending the answerer on a tirade of negativity and leading to hang-ups at comparable rates as when this question is not asked. Instead, I propose that lack of interest in and of itself can be a legitimate concern, and not necessarily a symptom of another concern; it should be treated as such and addressed directly. Still, statements of disinterest often indicate an impending hang-up and should be treated as red flags by interviewers.

On the other hand, some answerer utterances can be viewed as “green lights,” or signals that the answerer is open to participating. Questions about the length or content of the survey are more common in contacts where the answerer ultimately agrees to participate. Similarly, the presentation of conversation starters by answerers, such as questions directed at the interviewer or comments peripheral to the task at hand, are not only more frequent than concerns in agree contacts, but are much more frequent in agree compared to refusal contacts, and should be viewed as signs of engagement and likely participation.

Other research has looked at utterances by answerers that may indicate a greater likelihood of agreement. Work by Conrad et al. (2010), discussed in Section 4.3, found that answerers who use more backchannels such as “mm-

hmm” or “I see” seem to be indicating engagement in the conversation. These answerers are more likely to agree with the survey request than those who use fewer backchannels.

Being attuned to green lights in answerer speech, such as backchannels, questions about length or content, or the presentation of more conversation starters relative to the expression of concerns, can help interviewers know when an answerer is likely to agree and adapt their introduction in turn. In such a case, backing off from a “hard sell” may be recommended; however, it may also be advisable for interviewers to gently urge the answerer to begin the interview, rather than schedule a callback.

4.4 Suggestions for Future Research

Future research should be conducted to further this work and increase its applicability to different practice settings.

4.4.1 Rewording to Clarify Contrast between Listeners and Practitioners

Firstly, the contrast between listeners’ ratings of interviewer scriptedness and practitioners’ opinions about the importance of “following a script” should be revisited to ensure that this is an “apples to apples” comparison. As discussed in Section 2.4, practitioners may be thinking about scriptedness over the course of the whole introduction, possibly in terms of mentioning key points such as name, affiliation, and sponsor, while listeners are likely to be rating whether an interviewer sounds robotic or like she is reading. I would recommend repeating this research using a different word in the listeners’ study, such as “robotic,” and asking practitioners to rate the importance to an interviewer’s success of not only

following a script but also “sounding natural or not robotic in the initial seconds of an introduction.”

4.4.2 Exposure to Longer Clips of Interviewer Speech

Another expansion of the listeners’ study could include setting a higher threshold for inclusion of interviewer turns. As described in Section 2.2.2, the minimum amount of interviewer speech required for inclusion of a contact in this study was a statement of the interviewer’s name and affiliation. Replicating the listeners’ survey with longer excerpts of interviewer speech may show that characteristics which are not salient to judges during brief excerpts may actually be so, and may be associated with contact outcome, if they are rated based on a lengthier exposure. The listeners’ study used content from actual survey introductions, and only used the amount of interviewer speech that was uttered without any intervening speech from the answerer. To extend this, one could splice together multiple interviewer turns, while still removing any answerer speech. For example, a typical sequence in many contacts is:

Interviewer: Hello, this is ___ calling from the University of Michigan.

Answerer: Uh huh?

Interviewer: I’m calling about our nationwide study on the economy.

In the listeners’ study, judges were only exposed to the first interviewer turn, but a recommended extension would be to delete the answerer turn and present the two interviewer turns as one. Including answerer speech could influence judges to rate contacts based on the exchange (rather than the interviewer’s speech) or on how they think the answerer perceives the interviewer (rather than

how they themselves perceive her). However, omitting brief answerer turns, such as backchannels (as in the example above), could expand the amount of interviewer speech included.

4.4.3 Content Analyses of Interviewer Speech

An additional expansion of the listeners' study could include analyses of the content of the interviewer speech to which answerers are exposed, or at least controlling for this in models. This information is coded in the existing datasets, but was not explored at all in these analyses. It may be that mentioning the incentive in the initial move (which was done in 80 of 283, or 28 percent of contacts in the listeners' study) has more of an effect on response likelihood than judgments of interviewer characteristics. Further analyses on the same dataset used in the listeners' study could explore the order in which different factors (incentive, length, sponsor, etc.) are mentioned, and if there is any effect of earlier mentions of particular factors on response likelihood. Further, this type of data could offer an interesting comparison with the practitioners' study; practitioners rated as low the importance of "emphasizing the length of the survey" and "emphasizing the incentive" to an interviewer's success.

4.4.4 Interviewer-level Analyses

Replicating the listeners' study using a greater number of contacts by the same interviewers may shed light on those interviewer characteristics or behaviors across multiple contacts that lead to greater success. Conducting interviewer-level rather than contact-level analyses would necessitate a larger and differently structured corpus of contacts (for example, 50 contacts by each of

50 interviewers), but could better answer questions about best practices of interviewers.

4.4.5 Analyses of Variation in Ratings by Rater Characteristics

Another variation in design of the listeners' study could focus on how different raters judge the same contact. Results of this research could inform targeting of particular groups of respondents by interviewers with particular characteristics. For example, if it could be determined that men are more sensitive than women to an interviewer's scriptedness, then more experienced interviewers (who tend to be less scripted, as discussed in Section 2.3.4) could be assigned to call sample members who are known to be male. Obviously, this application is not feasible in a random-digit-dial survey, but could be considered when a list sample indicating gender is used.

Rater characteristics such as age, race, and education could be considered (implementing this would require a more diverse sample than the one used in the listeners' study). In addition, if both rater and interviewer age are known, analyses could explore how raters judge interviewers who are older than, younger than, or the same age as themselves. If there are differences, findings could be applied to targeted calling of sample members; for example, if it was known that female judges tend to rate male interviewers who are older than they are as competent and professional, and are more likely to respond to competent and professional interviewers, interviewers with these characteristics could be matched to sample members matching this profile. This may be most applicable in a list-based or longitudinal survey.

4.4.6 Analyses by Different Types of Scheduled Callbacks

An intriguing finding from the tailoring study is that interviewers' responsiveness scores are higher in scheduled callbacks than in cases of either agrees or refusals. This could indicate that by being highly responsive, interviewers are able to convert possible refusals to scheduled callbacks; however, further analyses of contacts that resulted in scheduled callbacks are recommended. Contacts are classified as scheduled callbacks if the interviewer states that she will call back, but there are several situations where this may occur: the interviewer and answerer can agree on a time for a subsequent call (a conversation which can be initiated by either the answerer or the interviewer); the interviewer can offer to call back as a response to a refusal by the answerer; or the interviewer can offer to call back to reach a specific household member who is not home at the time of the call. Repeating the analysis of responsiveness for each of these types of scheduled callbacks may reveal different levels of responsiveness in each category. Additionally, incorporating case-level data such as ultimate outcome and identifiers for all interviewers who had contact with the household can demonstrate differences among answerers who defer agreement.

Specific questions to be addressed could include:

- Is ultimate agreement more likely if an answerer initiates the scheduled callback (as opposed to an interviewer-initiated callback)?
- Is ultimate agreement more likely if an answerer is called back by the same interviewer, rather than a different interviewer? I would hypothesize that if a particular interviewer has been accommodating by allowing the

answerer to reschedule the interview at their convenience (rather than imposing at the first call), an answerer might feel obligated to talk to the same individual in a callback, but would not feel the same obligation to an unknown interviewer.

4.4.7 Distinguishing True from False Concerns

Another extension of the tailoring study discussed in Chapter 3 might look at the expression of concerns through the lens of Brown and Levinson's (1987) model of politeness. According to this model, answerers may express concerns in lieu of simply hanging up. By stating a concern, even if it is false, they are refraining from threatening the interviewer's "positive face," which Brown and Levinson define as "the positive consistent self-image . . . including the desire that this self-image be appreciated and approved of" (p. 323). Concerns may be genuine, but may also be an "off the record" way of expressing a lack of desire to participate. Using data on final case outcome, a hypothetical project could address the following research questions around distinguishing true from false concerns:

- Do paralinguistic elements such as intonation ("I don't have time" versus "I DON'T HAVE TIME!!!") or pauses before stating concerns offer any clues as to whether the case will eventually become an agree? One hypothesis might be that an answerer who pauses before offering, "Call me back tomorrow," is mentally reviewing his or her schedule, while one who does not pause is trying to end the call as quickly as possible.

- Does the content of answerer moves (concrete excuses such as “I’m having dinner” as opposed to generic concerns such as “It’s a bad time”) offer clues as to whether a concern expressed during a scheduled callback is “legitimate?” I hypothesize that answerers who take the time to explain their concerns are more likely to be willing to participate than those who do not.

4.4.8 Collecting Interviewer Ratings in a “Live” Situation

Finally, the possibility should be explored that perceptions of interviewers by Web survey respondents (judges) are vastly different from the perceptions of actual answerers to telephone survey requests. In a mixed-mode design, which would necessitate a rich frame containing both telephone numbers and email addresses, one could recontact both survey respondents and nonrespondents immediately after they have heard part or all of a telephone survey interviewers’ request, and ask for their participation in a brief Web survey in which they would rate interviewer characteristics. Comparisons of these data to those collected in the listeners’ study may show that characteristics are rated differently, and are differentially important, when contacts are heard in a Web survey compared to “live” on the telephone.

Obviously, it would be difficult to contact and persuade people who have just refused to participate in a traditional landline telephone survey to then participate in a Web survey. One remedy for this could be conducting the survey among mobile device users. After taking (or refusing to take) a brief survey on their mobile phone, sample members would receive a text message directing them to

take a Web survey. To increase the likelihood that nonrespondents to the phone survey would participate in the Web survey, a sizable incentive could be offered for participation in a short survey, which would result in high costs for this study. Another obvious limitation is that a sample of mobile device users is not directly comparable to the original respondents to the listeners' study, although 88 percent of Lightspeed panel members (of whom listeners' study respondents are representative) reported owning a mobile phone.

Appendices

Appendix 1: Details on Preliminary Studies

Preliminary Study 1

Cialdini's (1984) Principles of Compliance state that compliance with a request is more likely if the requester is authoritative (as opposed to unauthoritative) or likable (as opposed to unlikable). Groves, Couper, and Cialini (1992) attempted to link Cialdini's Principles of Compliance to the survey participation decision, but could find no evidence that these principles actually guide a potential respondent's decision. Work by van der Vaart et al. (2005) used factor analysis to derive overall interviewer approaches, which they called "authority" and "likability," from rated interviewer characteristics such as friendly, professional, and objective. However, a direct link between perceived authority/likability and survey participation is a gap in the literature. Therefore, a preliminary study was developed which intended to use ratings of interviewers' authority and likability as predictors of contact outcome.

Objectives and Hypotheses

The first aim of the preliminary study was to test the hypothesis that inter-rater reliability of judges' assessments of authority and likability could be attained. The literature indicates a precedent for reliable ratings of interviewer characteristics (Oksenberg and Cannell 1988; van der Vaart et al. 2005), and it was hypothesized that detailed definitions and instructions would result in reliable judgments of authority and likability. This preliminary study also had the aim of

exploring what personality traits or elements of interviewer speech drive judges' ratings of authority and likability.

Data and Methods

A preliminary data set of 1,215 contacts was stratified by outcome (agree, refuse, scheduled callback) and by presence or absence of a household listing (an often lengthy exchange which was excluded from transcription and coding). Contacts in the top and bottom length quintiles were excluded from the sampling frame. Additionally, contacts that were outliers in interviewer's rates of repairs (utterances such as "Sun-Monday"), restarts (utterances such as "to-tomorrow") (top 2 percentile), fillers (top 10 percentile) or pauses (top 1 percentile) were dropped, as were contacts with multiple answerers. After these criteria were applied to the corpus, 1,084 contacts remained. Contacts were randomly selected from each stratum; a total of 12 contacts were selected with varying outcomes (five agree, 5 refuse, 2 scheduled callbacks).

A convenience sample of twelve raters (6 males and 6 females, aged 26 to 66) was used. All rated each interviewer's authority and likability on a scale from 1 to 7 (full details provided in Appendix 3).

All raters listened to the same 12 contacts; however, to control for order effects, the order of presentations was varied, with half of the raters hearing one random order, and half hearing the contacts in reverse order. The design is summarized below.

Design of Preliminary Study 1

	Order of questioning	
Order of contact presentation	Likability-Authority	Authority-Likability
Order 1	3 raters	3 raters
Order 2	3 raters	3 raters

Following each contact, a brief, unstructured interview was conducted with questions probing the drivers of authority and likability ratings. Questions asked included: “Why did you choose that number?” “What else helped the interviewer’s authority/likability?” “You said the interviewer was _____. Did that help or hurt her authority or likability?” “What other words would you use to describe the interviewer?”

Findings

While ratings of broad concepts such as “authority” and “likability” were not reliable, qualitative findings were rich and informative as to the characteristics that influenced ratings of these concepts. Common dimensions were repeatedly mentioned as drivers of authority (e.g., competent) and likability (e.g., pleasant to listen to). The concept of tailoring or adaptability was also cited by judges as an important element in the introductory interaction.

When asked about the interviewer’s authority, respondents often discussed the idea of competence, or “understanding what was going on.”

“She stated the purpose of the study, she seemed to know what she was talking about.”

“He didn’t understand what was going on. There were pauses and a few uhhs; he sounded scared.”

Raters alluded to vocal attractiveness, or the idea that someone was “pleasant to listen to,” and some made links between attractive voices and their response propensity, stating that they “wouldn’t” or “couldn’t” listen to unattractive voices for an extended period of time.

“He had a nice voice. A nice-sounding voice. That’s really all.”

“Her voice was annoying. I wouldn’t want to listen to her for an hour. It was a nasally tone, and slow.”

Interviewers’ responsiveness or ability to adapt to answerers was a common theme, although it was not consistently associated with authority or likability, indicating that tailoring is indeed a critical feature of an interview request:

“He sounded like he would work with you. If you asked him something, he’d do it. That made him likable.”

“When the man made a request, he said he could accommodate it. That made him more authoritative—he had control over it, he could adjust what would happen.”

Findings from this study motivate the investigation of specific interviewer tactics, such as tailoring, and personality traits, rather than broader global concepts such as authority and likability. At the same time, these findings offer justification for the hypothesis that a warm and competent interviewer may garner responses, based on the Authority and Likability heuristics discussed by Cialdini (1984).

Preliminary Study 2

Objectives and Hypotheses

Two small Web surveys were conducted to test the feasibility of rating vocal attractiveness, and also to perform preliminary analyses testing the hypothesis that more attractive voices lead to longer contracts.

Data and Methods

Due to budget constraints, the design of these surveys was not elegant. Potential respondents received an email from the investigator which contained a link to a survey in Survey Monkey as well as an attached audio file, consisting of concatenated excerpts from actual contacts. These contacts were selected through stratified random sampling from a dataset of 496 contacts (all contacts by female interviewers with at least 10 contacts in the dataset, excluding contacts from the Gujarati Community Study but including those with only two interviewer turns), where strata were all contacts conducted by a particular interviewer. Excerpts from contacts were presented with five seconds of silence between them. The survey asked respondents to “please rate each voice's ATTRACTIVENESS on a scale from 1 (extremely unattractive) to 7 (extremely attractive).”

In the first survey, five audio excerpts of the first interviewer turn (typically along the lines of “Hello, my name is ____ and I’m calling from the University of Michigan about an important economic study”) were used. These were presented in two different orders, with half of respondents receiving the first order and half receiving the second order. Fifteen people responded to this survey.

In the second survey, 10 audio excerpts were used. Half of these contained the first use of “Hi” or “Hello” by the interviewer and half contained the interviewer saying “University of Michigan.” These were also presented in two different orders. Eleven people responded to this survey.

Findings

The first study, involving longer exposures but more variable content, resulted in agreement among respondents that two of the voices were clearly attractive and three were clearly unattractive. In the second study, although some contacts clearly fell on one side or the other of the scale, there was greater variance between respondents. For example, the same contact was rated as unattractive (1, 2, or 3 on a 7-point scale) by four respondents and attractive (5, 6, or 7) by four respondents. This result motivates the use of longer exposures (specifically, the first interviewer turn) in order to obtain greater inter-rater reliability.

In both studies, positive correlations were observed between the average attractiveness rating and the total number of conversational turns in the contact (.14 in the first survey and .28 in the second), as well as the number of answerer turns (.14 in the first survey and .27 in the second), offering support for the hypothesis that a more attractive voice can lead to longer contacts, but indicating that there may be an additional component driving length.

Preliminary Study 3

Objectives and Hypotheses

An additional Web survey was conducted to explore the hypothesis that length is driven not only by attractiveness, but also by competence.

Data and Methods

The design of this study was similar to the Web studies in Preliminary Study 2. Potential respondents received an email from the investigator which contained a link to a survey in Survey Monkey as well as an attached audio file, consisting of concatenated excerpts from actual contacts. These contacts were selected from a subset of contacts which included all contacts by female interviewers with at least 10 contacts in the dataset, excluding contacts from the Gujarati Community Study and those with only 2 interviewer turns. Contacts were selected based on length: four from the 25th percentile and four from the 75th percentile were included.

First turns from contacts were presented with five seconds of silence between them. The survey asked respondents to “please rate HOW COMPETENT the speaker sounds. Use a scale from 1 (not competent at all) to 6 (extremely competent).”

These were presented in two different orders, with half of respondents receiving the first order and half receiving the second order. Eight people responded to this survey.

Findings

While results were reliable across respondents (average weighted kappa of .23), the correlation between average competence rating and contact length was in the opposite direction as expected. Anecdotal feedback from respondents indicated that there was some uncertainty around the concept of “competent.”

For this reason, I plan to ask questions about multiple characteristics, including competent but also including “professional,” “confident,” and “knowledgeable.”

Appendix 2: Listeners' Survey

PROGRAMMING INSTRUCTIONS:

- EACH RESPONDENT RECEIVES 10 CLIPS: 5 IN PART 2 (LIKELIHOOD PREDICTION) AND 5 IN PART 3 (RATINGS). DO NOT USE THE SAME CLIP FOR BOTH PARTS FOR THE SAME RESPONDENT.
- EACH CLIP RECEIVES RATINGS FROM 60 RESPONDENTS: 30 ON PART 2 AND 30 ON PART 3.
- RANDOMIZE ORDER OF PRESENTATION OF CLIPS IN EACH PART (I.E., NO CLIP SHOULD CONSISTENTLY BE IN THE SAME POSITION).

I. SCREEN

1. What is your gender?
 - a. Male
 - b. Female

2. Which of the following categories includes your age?
 - a. Under 18 [TERMINATE]
 - b. 18-29
 - c. 30-39
 - d. 40-49
 - e. 50-59
 - f. 60+

3. Would you say that your ability to understand spoken English is...
 - a. Excellent
 - b. Good
 - c. Fair [TERMINATE]
 - d. Poor [TERMINATE]

QC. You will now hear a brief clip of someone speaking, similar to the clips you will hear during this survey. [PLAY CLIP]

Which of the following words did you hear? [RANDOMIZE. ALLOW ONE.]

1. Miniature
2. Nightmare
3. Michigan
4. Tennessee
5. Relative

PN: CONTINUE IF PRECODE 3 IS SELECTED.

*SET UP 5 BLOCKS OF RANDOMIZATION (UNDER 5SECS TO MORE THAN 15 SECS)

*SELECT 2 FROM EACH BLOCK USING A LEAST QUOTA SET UP.
EXAMPLE (RANDOMIZE ORDER):
Q4=<5SECS, 5-6.9SECS, 7-9.9SECS, 10-14.9SECS AND >15SECS
Q5=<5SECS, 5-6.9SECS, 7-9.9SECS, 10-14.9SECS AND >15SECS

II. LIKELIHOOD PREDICTION

4. You are about to hear the beginning of a few phone conversations. In each, the person speaking wants to convince the person on the other end of the phone to participate in a survey interview. You will only hear one person speaking for between 3 and 30 seconds. Just based on this, please give us your best guess as to HOW LIKELY it is that the person on the other end of the phone WILL AGREE TO PARTICIPATE in the survey.

You are free to listen to each recording as many times as you need to, although once might be enough!

[RESPONDENTS CAN HEAR MULTIPLE TIMES. CAPTURE NUMBER OF LISTENS FOR EACH CLIP. KEEP INTRO TEXT VISIBLE FOR ALL. PRESENT SCALE INDIVIDUALLY FOR EACH CONTACT, NOT AS GRID; RESPONDENTS SHOULD NOT BE ABLE TO SEE THEIR PRIOR RATINGS.]

Second screen: Here's the second phone conversation. Please give us your best guess as to HOW LIKELY it is that the person on the other end of the phone WILL AGREE TO PARTICIPATE in the survey.

You are free to listen to each recording as many times as you need to, although once might be enough!

Third screen: Here's the third phone conversation. Please give us your best guess as to HOW LIKELY it is that the person on the other end of the phone WILL AGREE TO PARTICIPATE in the survey.

You are free to listen to each recording as many times as you need to, although once might be enough!

Fourth screen: Here's the fourth phone conversation. Please give us your best guess as to HOW LIKELY it is that the person on the other end of the phone WILL AGREE TO PARTICIPATE in the survey.

You are free to listen to each recording as many times as you need to, although once might be enough!

Fifth screen: Here's the last phone conversation. Please give us your best guess as to HOW LIKELY it is that the person on the other end of the phone WILL AGREE TO PARTICIPATE in the survey.

You are free to listen to each recording as many times as you need to, although once might be enough!

	Not at all likely					Extremely likely
	1	2	3	4	5	6
How likely is it that the person on the other end of the phone will AGREE TO PARTICIPATE in a survey?						

III. RATINGS

5. You will now hear the beginning of a few more phone conversations. Again, the person speaking wants to convince the person on the other end of the phone to participate in a survey interview. And again, you will only hear one person speaking for between 3 and 30 seconds. Just based on this, please rate the speaker on the following characteristics. You are free to listen to each recording as many times as you need to, although once might be enough!

Second screen: Here's the second phone conversation. Just based on this, please rate the speaker on the following characteristics. You are free to listen to each recording as many times as you need to, although once might be enough!

Third screen: Here's the third phone conversation. Just based on this, please rate the speaker on the following characteristics. You are free to listen to each recording as many times as you need to, although once might be enough!

Fourth screen: Here's the fourth phone conversation. Just based on this, please rate the speaker on the following characteristics. You are free to listen to each recording as many times as you need to, although once might be enough!

Fifth screen: Here's the last phone conversation. Just based on this, please rate the speaker on the following characteristics. You are free to listen to each recording as many times as you need to, although once might be enough!

[ROTATE ATTRIBUTES. ASK ALL FOR EACH OF 5 CLIPS.]	Not at all					Extremely
	1	2	3	4	5	6
Confident						
Professional						
Pleasant to listen to						

Competent						
Knowledgeable						
Natural-sounding						
Enthusiastic						
Genuine						
Scripted						
Friendly						
Uncertain						
Irritating						

IV. DEMOGRAPHICS

We just have a few more questions for classification purposes.

D1. In which state do you currently reside?

AK	ID	MT	RI
AL	IL	NE	TN
AR	IN	NV	TX
AZ	KS	NH	UT
CA	KY	NJ	VA
CO	LA	NM	VT
CT	ME	ND	SD
DC	MD	NC	SC
DE	MA	NY	WA
FL	MI	OH	WI
GA	MN	OK	WV
IA	MO	OR	WY
HI	MS	PA	

NORTHEAST

Maine (ME), New Hampshire (NH), Vermont (VT), Massachusetts (MA), Rhode Island (RI), Connecticut (CT), New York (NY), New Jersey (NJ), Pennsylvania (PA)

MIDWEST

Ohio (OH), Indiana (IN), Illinois (IL), Michigan (MI), Wisconsin (WI), Minnesota (MN), Iowa (IA), Missouri (MO), North Dakota (ND), South Dakota (SD), Nebraska (NE), Kansas (KS)

SOUTH

Delaware (DE), Maryland (MD), Washington DC, Virginia (VA), West Virginia (WV), North Carolina (NC), South Carolina (SC), Georgia (GA), Florida (FL)

Kentucky (KY), Tennessee (TN), Alabama (AL), Mississippi (MS), Arkansas (AR), Louisiana (LA), Oklahoma (OK), Texas (TX)

WEST

Montana (MT), Idaho (ID), Wyoming (WY), Colorado (CO), New Mexico (NM), Arizona (AZ), Utah (UT), Nevada (NV)

Washington (WA), Oregon (OR), California (CA), Alaska (AK), Hawaii (HI)

D2. Which of the following best describes your ethnicity or race?

- White
- Black or African-American
- Asian
- American Indian or Alaskan Native
- Other ethnic group
- Prefer not to answer

D3. Are you of Hispanic or Latino origin?

- c. Yes
- d. No

D4. Into which of the following categories does your 2010 household income fall?

- a. Less than \$30,000
- b. \$30,000 to \$39,999
- c. \$40,000 to \$49,999
- d. \$50,000 to \$59,999
- e. \$60,000 to \$69,999
- f. \$70,000 to \$79,999
- g. \$80,000 to \$89,999
- h. \$90,000 to \$99,999
- i. \$100,000 or more
- j. Prefer not to answer

D5. What is the highest level of education you have completed?

- a. Less than high school
- b. High school diploma or GED
- c. Some college
- d. College degree
- e. Some graduate school
- f. Graduate degree

Thank you for your participation!

Appendix 3: Sampling Structure

For each study in the sampling frame, a sample of households, referred to as “cases,” was selected. The sampling frame consisted of all contacts from the five selected studies (558,695 contacts in total). To select the cases included in the dataset, the investigators stratified the cases first by interviewer and then by the outcome of the case’s first contact. For purposes of stratification by interviewer, cases were assigned to the interviewer who made the first contact on the case; 100 interviewers have first contacts. For each interviewer, up to eighty cases were sampled: up to forty cases where the first contact was positive (agreement) and up to forty cases where the first contact was negative (refusal or “other outcome”). If an interviewer had fewer than forty cases in either stratum, all cases in that stratum for that interviewer were included. For each sampled case, all contacts in the case were selected; however, approximately 30 percent of sampled contacts were not included in the dataset due to random recording failures.

Forty-one replicates were created. Each of the first forty replicates had all contacts from two cases (one with a positive first contact and one with a negative first contact) assigned to each interviewer. An additional replicate, referred to as the “take-all” replicate, included all contacts from cases where the first contact was by an interviewer who had fewer than forty positive or negative first contacts. Cases were assigned to replicates using a systematic selection of cases with a single random start within each interviewer set. All contacts from the “take-all” replicate and the first four regular replicates were transcribed, coded, and rated.

Appendix 4: Practitioners' Survey

I. SCREEN

1. In which of the following areas do you have decision-making authority?
[SELECT MULTIPLE UNLESS C]

- a. Interviewer hiring
- b. Interviewer training
- c. Neither of these

[TERMINATE IF C TO Q1]

II. ORGANIZATION INFORMATION

2. How many CATI stations does your organization have? [NUMERIC TEXT BOX. INCLUDE NOT SURE OPTION.]

3. What percentage of the telephone interviews conducted by your organization are for [A, B, C MUST SUM TO 100% IF D NOT SELECTED]

- a. Government, nonprofit, or academic organizations
- b. For-profit organizations
- c. Other
- d. Not sure

4. Approximately how many telephone interviews did your organization complete during the 2010 calendar year? [SELECT ONE]

- a. Fewer than 1,000
- b. 1,000–4,999
- c. 5,000–9,999
- d. 10,000 or more
- e. Not sure

III. WHAT MAKES A GOOD INTERVIEWER?

5. In your opinion, how important is each of the following to an interviewer's response rate, that is, success in obtaining interviews? [ROTATE. SELECT ONE PER ROW.]

	Extremely important	Somewhat important	Not very important	Not at all important
A pleasant-sounding voice				
The ability to convey knowledge about the study				
Professional demeanor when talking to potential respondents				
The ability to address relevant concerns expressed by potential respondents				
The initial impression an interviewer gives to sample members				
The interviewer's speech rate				
How high or low the interviewer's voice sounds (pitch)				
An interviewer's voice that does not sound monotonous (has pitch variability)				
The interviewer's ability to follow a script during an introduction				
The interviewer's ability to "ad lib" or deviate from a script during an introduction				
Frequency of disfluencies, such as "um" and "uh"				
How confident the interviewer sounds to potential respondents				
How competent the interviewer sounds to potential respondents				
How enthusiastic the interviewer sounds to potential respondents				

How friendly the interviewer sounds to potential respondents				
How genuine the interviewer sounds to potential respondents				

6. Are there other elements not mentioned above that you think are important to an interviewer's response rate? What are they? [OPEN ENDED]

IV. HIRING INTERVIEWERS [ASK THIS SECTION ONLY IF Q1=A]

7. When you are HIRING interviewers, how important are each of the following criteria to your hiring decisions? [ROTATE. SELECT ONE PER ROW.]

	Extremely important	Somewhat important	Not very important	Not at all important
English language fluency				
Literacy level				
Ability to multitask				
Performance in training exercises				
Speaking voice				
Personality				
Prior experience				
References from previous employers				
Candidate seems reliable				
Candidate seems trainable				

8. Are there other elements not mentioned above that are important criteria to your hiring decisions? What are they? [OPEN ENDED]

9. [IF SPEAKING VOICE IS SOMEWHAT/EXTREMELY IMPORTANT IN Q7] You indicated that a candidate's SPEAKING VOICE was [SOMEWHAT/EXTREMELY] important to your HIRING DECISIONS. How important are each of the following elements of a candidate's voice to your hiring decisions? [ROTATE. SELECT ONE PER ROW.]

	Extremely important	Somewhat important	Not very important	Not at all important
Rate of speaking				
Pitch of voice, that is, how high or low the voice sounds				
Volume of voice				
Voice conveys enthusiasm				
Voice conveys confidence				
Voice conveys knowledge				
Voice sounds professional				
Voice is pleasant to listen to				
Voice sounds friendly				

10.[IF SPEAKING VOICE IS SOMEWHAT/ EXTREMELY IMPORTANT IN Q7]
Are there other elements of an interviewer's speaking voice not listed above that are important to your hiring decisions? What? [OPEN ENDED]

11. Have you ever, even once, NOT HIRED a candidate for a telephone interviewer position because of their voice? [SELECT MULTIPLE.]

- a. Yes
- b. No

12.[IF Q11=A] You indicated that you have had the experience of NOT HIRING a candidate for a telephone interviewer position because of their voice. What was it about the voice that kept you from hiring the person?

V. TRAINING INTERVIEWERS [ASK THIS SECTION ONLY IF Q1=B]

13. For each of the following, please indicate if it is a primary focus, a secondary focus, or not a focus at all in telephone interviewer training. [KEEP 1 AND 2 TOGETHER. ROTATE.]

	Primary focus	Secondary focus	Not a focus
Obtaining interviews in general			
Obtaining interviews from reluctant respondents (including refusal conversion)			
Standardized interviewing skills			
Use of CATI system			

Presenting a confident demeanor			
Responding appropriately to sample members' arguments or concerns			
Developing knowledge about the study			
Being able to answer questions about the study or organization			
Importance of data quality			
Following introductory scripts			
Developing a personalized or non-scripted introduction			
Recruiting respondents			
Administering interviews			

14. Are there other elements of interviewer training that are PRIMARY FOCUSES in your organization? What are they? [OPEN ENDED]

Appendix 5: Email Request Sent to Practitioners

Dear Colleague,

My name is Jessica Broome and I am a PhD candidate in Survey Methodology at the University of Michigan. I am conducting research on telephone interviewers and would love to hear the thoughts of **an individual in your organization responsible for hiring and training telephone interviewers**. I hope you can contribute 15 or 20 minutes of your time to complete a brief survey. It can be found at http://umichisr.qualtrics.com/SE/?SID=SV_3wJBdT1iLeNGuQ4. If someone other than you is better suited to answer questions about hiring and training telephone interviewers in your organization, I hope you will forward this message to him or her.

Responses will never be linked to the identity of the respondent or their organization! However, if you would like to receive a copy of the results, please provide your contact information at the end of the survey.

If you have any questions, please feel free to contact me at jsbroome@umich.edu or my advisor, Dr. Frederick Conrad, at fconrad@isr.umich.edu.

My sincere thanks for your time and participation.

Warmly,
Jessica

Jessica Broome
PhD Candidate
Michigan Program in Survey Methodology

NOTE: This email was sent to all members of the sampling frame using the blind copy function. However, because too many characters were entered into the blind copy field, most recipients received the message multiple times (reports indicated some respondents receiving as many as 93 copies of the survey invitation).

Thirty-seven recipients emailed the investigator, with messages ranging from concerned to irate. That day and the next, individual apologies were sent to all message recipients individually, explaining the situation and urging them to consider participating. Of these, five emails failed; three people responded and explained that they had no research staff; and one responded that he was not interested.

Appendix 6: Details on Reliability of Initial Move Coding

Two subsets of twenty contacts (audio recordings and transcripts) were randomly selected for a reliability test among the five coders who coded more than 90 percent of the contacts. Because decomposing turns into moves is itself a judgment about which there can be disagreement, we tested two subsets of contacts in order to measure agreement with and without the “turn decomposition” task. One subset of contacts was presented to the coders with transcripts in their original form, that is, not broken into moves. A second subset of twenty contacts for which the transcripts had already been broken into moves was coded by all five coders and tested for reliability. Each of the five coders decomposed the turns in the first subset of twenty contacts into moves and assigned codes to each move. Weighted kappa, as opposed to simple kappa, was used to measure reliability because with the multivariate coding system, the chances of perfect agreement on 28 variables is very remote; weighted kappa takes into account the complexity of the coding structure and essentially gives “partial credit.” Weighted kappa for each pair of coders ranged from .53 to .93, mean = .74. For the second set of twenty contacts (already structured by moves) weighted kappas ranged from .77 to .86, mean = .82. We interpret these scores as indicating strong agreement among coders.

Appendix 7: Move Codes

Answerer:

- 1: Expression of enthusiasm (I was waiting for your call).
- 2: Expression of interest.
- 3: Acknowledge interviewer (I know it's you).
- 4: Request clarification.
- 5: Suspicion or misunderstanding of purpose (I don't want to buy anything, take me off your list).
- 6: Past experience with research (I did a survey last week).
- 7: UM-related.
- 8: Expression of disinterest.
- A: Statement of ambivalence.
- B: General confirmation (Yes, that's right).
- C: Contact information/ confirmation (This is a home).
- D: Duration question (How long will this take?).
- E: Duration comment.
- F: Follow-up comment (My wife said you called).
- G: Formal greeting (Hello).
- H: Answer to household roster question.
- I: Self-Identification (This is Mike).
- J: Informal greeting.
- K: Follow-up question (Didn't you call before?).
- L: Procedural question (How does the randomization work?).
- M: Incentive question.
- N: Not interested but not explicit refusal.
- O: Other.
- P: Willing to participate.
- Q: Procedural comment.
- R: Explicit refusal.
- S: Scheduling statement (10:00 would be fine).
- T: Not enough time.
- U: Backchannel.
- V: Incentive comment.
- X: Scheduling question.
- W: Personal conversation—non-interview related.
- Y: Request for identification (Who is this?).
- Z: Eligibility-related (This is not my home).

Interviewer:

- 9: first question of interview.
- a: indirect invitation.
- b: direct invitation.
- c: contact information/ confirmation (Have I reached you at . . .).
- d: description of survey (This survey helps us understand what people think about the US economy).
- e: why not participate.
- f: follow-up: any (reference past call, other household member, or object).
- g: formal greeting.
- h: household listing statement/question (How many adults live in this household? The computer is going to choose the person I need to speak with).
- i: self-identification.
- j: informal greeting.
- k: duration comment.
- l: apology.
- m: incentive-related.
- n: expression of gratitude.
- o: other.
- p: persuasive statement (Your opinions are very important to us.).
- q: establishing legitimacy (You can visit our Web site).
- r: procedural information (This call may be recorded for quality control).
- s: scheduling statement (Is there a better time?).
- t: break into pieces (We can just do a few questions).
- u: backchannel.
- v: general confirmation.
- w: sweetening the pot (Now I can offer you \$20 . . .).
- x: personal conversation—non-interview related.
- y: acknowledge imposition (I know it is late/ It sounds like you're busy).
- z: scarcity reference (There are only two days left).

Appendix 8: Tailoring Study: Coding Scheme

Variable: TailorMove

Answerer codes:

O: no tailoring or tailoring opportunity.

A: concern: time.

B: concern: incentive.

C: concern: don't know anything / not suitable respondent (I don't have insurance/ I don't know anything about that).

D: concern: privacy/confidentiality.

E: concern: hearing.

F: concern: sick.

G: concern: other person needs to be involved in decision.

H: concern: don't understand.

I: concern: too fast.

J: concern: letter.

K: concern: not interested.

L: concern: do not call list/ take me off list (includes How did you get this number).

M: concern: don't want to buy anything.

N: concern: purpose/ content.

O: concern: other.

P: concern: other person (not interested/ not available).

Q: concern: length.

R: concern: household listing.

S: conversation starter.

T: concern: why was this number selected.

U: concern: harassment.

Interviewer codes:

- a: address concern: time.
- b: address concern: incentive.
- c: address concern: don't know anything/ not suitable.
- d: address concern: privacy/ confidentiality.
- e: address concern: hearing (INCLUDES PERCEIVED INCREASE IN VOLUME).
- f: address concern: sick.
- g: address concern: other person needs to be involved in decision.
- h: address concern: don' t understand.
- i: address concern: too fast (INCLUDE JUDGED CHANGE IN SPEED).
- j: address concern letter.
- k: address concern not interested.
- l: address concern: do not call list/ take me off list.
- m: address concern: don't want to buy anything.
- n: address concern: purpose/ content.
- o: address concern: other.
- p: address concern: other person (not available/ not interested).
- q: address concern: length.
- r: address concern: hhl.
- s: reference earlier call with specifics.
- t: response to A question or comment.
- u: answerer-centric remark (address A by name; is this a good time).
- v: acknowledge concern without addressing (oh absolutely or yes i see).
- w: address concern: why household selected.
- x: address concern: harassment.

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