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A New Version of the Item Count Technique for Asking Sensitive Questions: Testing the Performance of the Person Count Technique

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Abstract

This paper presents empirical evidence on a recent advancement of the item count technique (ICT, a survey technique for asking sensitive questions), namely, the person count technique (PCT; Grant, Moon, & Gleason, 2014). PCT utilizes person lists instead of lists of filler questions, as is the case in the classic ICT design. This simplifies the questioning procedure, but leads to some methodological challenges such as floor and ceiling effects. The main part of this paper presents empirical evidence stemming from an experimental postal survey in Germany ($N = 580$) investigating how well PCT performs as compared to standard direct questioning (DQ) with regard to alleviating misreporting for questions on attitudes towards refugees.

PCT prevalence estimates for hostile attitudes towards refugees are significantly higher than DQ estimates for one item, and non-significantly higher for three items. Although not consistently significant, the differences are substantial, amounting to a threefold increase of the proportion of respondents expressing negative attitudes towards refugees. Even though the findings are not unequivocally in favor of PCT, this new ICT variant still deserves consideration in the future and warrants further development. Specifically, more knowledge is required with respect to its statistical properties and the best practices of its implementation.

Keywords: Sensitive questions, response bias, misreporting, item count technique, person count technique, refugees, xenophobia.



Background and Research Question

The issue of so-called sensitive questions has occupied survey methodology for several decades (Barton 1958; Hyman 1944; Krumpal 2013; Tourangeau & Yan 2007). It is a well-established fact that respondents, when answering survey questions on socially undesirable or desirable behaviors or attitudes, tend to tailor their answers in a socially desirable manner rather than answering truthfully. This pertains to questions on socially loaded behaviors (e.g., self-reported delinquency, voting behavior, or substance abuse), attitudes (e.g., xenophobia or homophobia), as well as other personal traits (e.g., health issues or personality characteristics). Generally speaking and according to Tourangeau and Yan (2007, p. 860), sensitive questions in surveys can be defined as questions which are private or intrusive, which pose a threat of disclosure for the respondent, and/or touch upon socially undesirable or desirable topics. The primary problem of misreporting on such questions by respondents in standard survey settings is that prevalence estimates of sensitive behaviors or attitudes will be biased. For example, Bradburn and Sudman (1979, p. 24) compare survey estimates of self-reported alcohol consumption with official sales figures, finding that “reported beer, wine, and liquor consumption [...] reaches only 51, 67, and 36 percent of the taxed sales figures, respectively”. Furthermore, correlations between the sensitive issue under investigation and its determinants are also biased if the likelihood of misreporting is related to the determinants (Ganster, Hennessey, & Luthans 1983). Yet another issue when asking sensitive questions in surveys is item-nonresponse, which occurs if respondents refuse to answer the respective question at all. While this is a well-known phenomenon concerning questions on income (Moore, Stinson, & Welniak 2000; Yan, Curtin, & Jans 2010), empirical evidence is less consistent for sensitive questions on other topics

Editorial note

The author of this article is one of the guest editors of this MDA special issue. To avoid conflict of interest, the article has been handled by the other two guest editors and the author has not been involved in any of the editorial work related to this article (such as selecting and inviting reviewers, deciding on revisions, and accepting the manuscript for publication). The identities of the reviewers have not been revealed to the author.

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(Tourangeau & Yan 2007, p. 862). This could be because respondents may interpret an answer refusal as an “admission of guilt”.

In order to tackle the problem of misreporting (and item-nonresponse), survey methodologists have come up with a number of special questioning techniques. Conventional approaches encompass, for instance, anonymity assurances, “forgiving wording”, or the sealed envelope technique (Benson 1941; Perry 1979). A more elaborate procedure is the randomized response technique (RRT; Fox & Tracy 1986; Warner 1965), which has probably gained the most attention in the methodological literature on sensitive questions in surveys. However, RRT procedures in surveys are usually complicated both for respondents and for interviewers. Moreover, doubts have been raised regarding the efficacy of RRT in avoiding response biases (Wolter & Preisendörfer 2013). An alternative to RRT is the item count technique (ICT, also referred to as list experiment or unmatched count technique; Droitcour et al. 1991; Kuklinski, Cobb, & Gilens 1997), which has attracted increased interest within the research community in recent years.

As with RRT, the idea behind ICT is the anonymization of the interview situation by adding noise to the data concealing the respondents’ answers. This is achieved by randomly splitting the sample into (at least) two groups. One group, the “short-list group”, receives a list of binary yes-no questions which are “harmless” and function as filler items (i.e., they are not important with regard to their content). The other group, the “long-list group”, receives the same list of non-key items, but this time, the list additionally contains the (binary) sensitive item of interest. Respondents in both groups are asked not to answer each item individually, but rather to merely report the number of “yes” answers to the whole list. Therefore, the individual answer to the sensitive item is not disclosed to anyone, not even the interviewer (unless ceiling or floor effects occur, see below). For the whole sample, however, it is possible to calculate an estimate of the prevalence of the sensitive item by simply subtracting the mean of the short list from the mean of the long list. This classic ICT design for binary yes-no questions has recently been expanded upon with a version called item sum technique (IST; Trappmann et al. 2014; Wolter & Herold 2018), designed for quantitative sensitive items (such as the frequency of drug usage).

The person count technique (PCT) is another new variant of the classic ICT approach, originally proposed by Grant et al. (2014). PCT also applies to binary sensitive items, but instead of using lists of filler questions, it utilizes lists of persons. The short list is a number of people, and respondents are asked to report the number of persons for whom something applies. The long list corresponds to a list of persons as well, but also contains the respondent himself or herself.

This study presents empirical evidence on the performance of PCT as compared to standard direct questioning (DQ) with regard to alleviating misreporting on sensitive questions. To my knowledge, apart from the original (unpublished)

study by Grant et al. (2014), there exists, as yet, no published research investigating the performance of the only just recently proposed PCT. The empirical data presented here were gathered in a postal survey of $N = 580$ respondents in the City of Mainz, Germany. The PCT-DQ comparison is investigated for four questions on attitudes towards refugees/asylum seekers in Germany. According to the literature (Krumpal 2012; Stocké 2007), expressing negative or hostile attitudes towards immigrants is prone to underreporting. Therefore, due to the enhanced anonymity in PCT mode as compared to DQ, self-reports on hostile attitudes towards refugees should be higher in PCT mode as compared to DQ mode (and, if item-nonresponse is a problem, it should be lower in PCT mode than in DQ mode).

The structure of this article is as follows: The next section will give a brief overview of methodological research on response biases pertaining to attitude questions about immigrants. Afterwards, I will first present the principles of ICT and PCT in more detail, followed by a discussion of methodological aspects and some general pros and cons of PCT vis-à-vis ICT. The “Study Design and Methods” section is devoted to the description of the survey design and some issues of the statistical analyses. The “Results” section depicts the results regarding the PCT-DQ comparison, which are subsequently discussed within a broader framework in the final “Discussion” section.

Social Desirability Bias in Research about Xenophobia

There is a long tradition of research on anti-immigrant or xenophobic attitudes in the social sciences (Allport 1954; Czymara & Schmidt-Catran 2016; Quillian 1995; Weins 2011, to cite but a few). One of the motivations driving this literature is the public and scientific concern regarding political extremism, or, more specifically, regarding voting for (right wing) extremist parties in elections, for which anti-immigrant attitudes are seen as a major influencing factor (Arzheimer 2008). Studying the causes and consequences of xenophobia, however, requires a valid measurement of these attitudes. Several authors have argued that survey estimates from questions on anti-immigrant attitudes are prone to social desirability bias (An 2015; Cea D’Ancona 2014; Janus 2010; Krumpal 2012; Stocké 2007). Since there are social norms inhibiting the public utterance of such attitudes or opinions, some respondents may seek to avoid expressing them in survey interviews. This leads to the underreporting and underestimation of xenophobic attitudes.

In contrast to other (behavioral) sensitive issues, studying misreporting on attitude questions such as on xenophobia is not straightforward with respect to the level of response bias, because a “true value” cannot be observed (by using external validation records, for instance). Hence, in order to assess the amount of social

desirability bias, existing studies concentrate on comparing varying estimates according to different questioning techniques or survey modes. The ensuing evaluation is then carried out relying on the “more is better” assumption, which means that for socially undesirable traits like anti-immigrant attitudes, higher estimates are considered to be more valid than lower ones.

There are three studies comparing DQ and RRT estimates. Krumpal (2012) finds a significant improvement due to RRT for one out of three items on xenophobia, the prevalence of respondents expressing a xenophobic attitude amounting to 27 percent in DQ mode and to 35 percent in RRT mode. The estimates for the remaining two items are virtually the same in both question formats and amount to about 40 and 30 percent, respectively. Ostapczuk, Musch, and Moshagen (2009) observe a non-significant difference between a DQ and an RRT question on xenophobia. Depending on the education level of the respondents, the prevalence estimates of expressing a xenophobic attitude range from 25 to 45 percent in DQ mode and from 47 to 76 percent in RRT mode. Finally, Hoffmann and Musch (2016) compare the crosswise-RRT, an adjusted version of RRT (Yu, Tian, & Tang 2008), with DQ for one item on xenophobia and one on islamophobia. They observe significantly higher estimates (49 versus 27 percent) using the crosswise model for the first item, but not for the second (52 versus 43 percent).

Studies investigating the effect of ICT on self-reports of anti-immigrant attitudes have also been conducted. An (2015) finds that, when asked directly, around 59 percent (depending on education) of the respondents are against “cutting off immigration to the United States”. When asked using ICT, this fraction shrinks significantly to around 33 percent. Significant differences between DQ and ICT have also been reported by Janus (2010) for the same item (58 vs. 39 percent), and by Cappelen and Midtbø (2016) for an item on welfare benefits for immigrants in Norway. The study by Creighton and Jamal (2015), in contrast, yields mixed results with respect to the DQ-ICT comparison. While there is no difference for an item on “granting citizenship to a legal immigrant who is Muslim”, a significant difference (28 vs. 11 percent) was observed for “granting citizenship to a legal immigrant who is Christian”.

In sum, empirical research clearly shows that survey questions on anti-immigrant or xenophobic attitudes suffer from social desirability bias. The evidence regarding the performance of special survey techniques such as RRT or ICT to alleviate this problem, however, is mixed. The remainder of this article will present evidence on the performance of PCT in this regard.

Person Count: A Recent Advancement of the Item Count Technique

As explained above, the basic idea of ICT and PCT lies in concealing respondents' answers to binary sensitive survey questions by overlaying the data with noise. This noise is created by adding information about respondents' answers to other filler items (ICT) or third persons (PCT) to the individual answer to the sensitive item. Both ICT and PCT require a random split of the sample into a short-list group and a long-list group. When using ICT, respondents in the short-list group receive a list of harmless yes-no items, for example (Wolter & Laier 2014): "Below you see a list of four questions. Please indicate only the number of questions you answer with 'yes', thus, a number between zero and four. 1. Have you ever been abroad? 2. Have you ever used a taxi? 3. Have you used a plane this week? 4. Did you wash your car this week?". Respondents in the long-list group receive a list containing the same four non-key items plus the sensitive item of interest, for example "Have you ever driven a car while drunk?". Again, respondents are asked to only report the number of items they answer with "yes". In doing so, the individual answer to the sensitive item is not disclosed. Of course, this is true only if no ceiling or floor effects occur, i.e., if the respondent does not negate all items in the list or reports that all items apply. In order to avoid ceiling and floor effects, the non-key items should contain both low-prevalence and high-prevalence questions which ideally are negatively correlated among each other (Droitcour et al. 1991).

The PCT replaces the list of filler questions with a list of persons, and respondents are asked to report the number of persons for whom (they think that) something (sensitive) applies. In the short-list group, the list only consists of other uninvolved people; in the long-list group the respondent himself is added to the list; respondents report the number of persons for whom something applies including themselves. In the original proposition by Grant et al. (2014, p. 11–12) respondents were asked the following question: "We want to know what type of candidates people would support for President of the United States. Because this is a sensitive topic, we are not going to single you out. Instead, please think about three people you see or talk to often and we're going to ask you how many of these three people might be willing to vote for each type of candidate. We're going to ask about five candidates: a Republican, a Democrat, a Tea Party candidate, a Mormon, and a woman. It's ok to guess if you are not sure how many of the three people would vote for each candidate. [...]" In the short-list group the introduction subsequently read "Thinking of these three people, how many would be willing to vote for [a republican, a democrat, a woman etc.]", while in the long-list group, it read "Thinking of you and these three people [...]"

For both the basic ICT and the PCT, a prevalence estimate of the sensitive item $\hat{\pi}$ and its standard error can be calculated using the formulae (1) and (2) below,

provided that the short-list and long-list samples are independent. \bar{x}_{LL} and \bar{x}_{SL} represent the mean of the reported numbers in the long-list and short-list group, and $Var(\bar{x})$ the sampling variance of the mean estimate.

$$\hat{\pi} = \bar{x}_{LL} - \bar{x}_{SL} \quad (1)$$

$$S.E. (\hat{\pi}) = \sqrt{Var(\bar{x}_{LL}) + Var(\bar{x}_{SL})} \quad (2)$$

One advantage of the PCT design vis-à-vis ICT is that having one list of persons means that many sensitive items can be asked at once in the same survey. With ICT, a different item list is required for every sensitive item due to anonymity concerns (or, as Grant et al. 2014, p. 6, put it, an additional random split of the sample for every additional sensitive question, when using the same short list for every item). Also, no fabrication of artificial filler items is necessary with PCT, which could, in turn, simplify the answering process for the interviewees because they only have to deal with one question instead of a question list. But this, of course, has to be investigated empirically. One should also note that respondents may not be certain whether the trait being asked about applies to one or more of the uninvolved persons in the list. As cited above, Grant et al. (2014) try to solve this problem by prompting respondents “to guess if you are not sure”. If the interviewees follow this instruction, possible errors in judging about the status of the “other persons” represent no problem for the validity of the PCT estimate because, due to the experimental design (random split into short-list and long-list), the errors in both groups will be equal (Grant et al. 2014, p. 19) – provided that there are no design effects (see below). There are, however, some other challenges inherent to the PCT design, namely floor and ceiling effects, statistical power issues, and design effects. These challenges share (at least to some extent) a common cause, namely homophily effects, which I shall discuss first.

Homophily refers to the “similarity between socially connected individuals” (Shakya, Christakis, & Fowler 2017, p. 158). It is a well-established fact that similar people have a higher tendency to be socially connected than dissimilar people. This applies with respect to a variety of socio-demographic, behavioral, and attitudinal characteristics (McPherson, Smith-Lovin & Cook, 2001), including possibly sensitive traits such as marijuana consumption, political orientation, and delinquency (Kandel 1978; South & Felson 1990). One consequence of homophily regarding PCT is that it will affect the composition and characteristics of the “other persons”: When asked to think of some people whom they know, respondents probably unconsciously choose people who are similar to themselves, or at least more similar than a random choice would be. As the cited literature shows, this will also hold for the sensitive traits being asked about in the PCT procedure. Another, related argu-

ment is that the choice of the “other persons” may be guided by the question content and context (certain stimuli make respondents think of certain types of people). For instance, if the survey question deals with substance abuse, a respondent who smokes marihuana is probably going to imagine a list of “other persons” who are also inclined to smoke marihuana. This conjecture is supported by empirical evidence from social network research on name generators, which shows that question content and context exert an influence on the data generated by name generators in survey settings (e.g., Ferligoj & Hlebec 1999; Shakya et al. 2017). One finding of this research is also that individuals have different networks for different issues: “A person with whom someone discusses politics may not be the person upon whom they rely for assistance with a sick child” (Shakya et al. 2017, p. 158). A third argument for the occurrence of homophily effects (referring to values or attitudes) in PCT designs is derived from research showing that actors often subjectively overestimate the degree to which their acquaintances are similar to them (Huckfeldt & Sprague 1995): “People tend to assume that their friends are like them, when in fact areas of disagreement simply are not discussed” (McPherson et al. 2001, p. 429). Hence, when asked about characteristics of their acquaintances in PCT procedures, respondents may ascribe similar traits to the “other persons” even if this is objectively not the case.

In short, when using PCT we should expect that respondents generate lists of uninvolved persons that, due to homophily, share similar characteristics as themselves. This is probably further reinforced by framing effects of the question content and context, and by a subjectively overestimated degree of similarity by the respondents.

A first consequence of homophily effects with respect to PCT concerns floor and ceiling effects. As already pointed out, floor and ceiling effects occur if respondents either deny or affirm all items (persons) in the list. In this case, the anonymity of the procedure is negated. When using ICT, this can be avoided by a proper choice of the non-key items (negatively correlated high- and low-prevalence items), which is generally under the control of the researcher. When using PCT, floor and ceiling effects are likely to occur more often than with ICT because of homophily. Moreover, they are not as easily controllable as in the basic ICT design, because the choice of the uninvolved persons is not under the control of the researcher – at least in the PCT version proposed by Grant et al. (2014; see the discussion section below for a suggestion of how to possibly advance with this issue). My – preliminary – suggestion regarding the problem of floor and ceiling effects in the PCT design is to instruct respondents in a way that induces them to choose “other persons” that are as different as possible, and to carefully study floor and ceiling issues empirically both in the pretest phase of the survey and with respect to its main results. Also, one should take care not to introduce PCT as a “completely anonymizing technique” to

respondents. If floor and ceiling effects occur, respondents may feel cheated by the survey authors.

Another consequence of homophily effects, directly related to the issue of floor and ceiling effects, are issues of statistical power. One main drawback of all ICT designs is that they always produce larger standard errors than conventional estimates. This is obvious, because noise is artificially added to the data. The amount and the statistical properties of this noise affect the statistical efficiency of ICT estimates, which means that design aspects of the ICT/PCT procedure affect statistical efficiency and that there is a trade-off between efficiency and respondent protection (Coutts & Jann 2011; Trappmann et al. 2014). The standard errors of ICT estimates depend on (among other things) the number of non-key items (or the number of uninvolved persons in the PCT procedure), their prevalence, and the covariance between the sensitive item and the filler items (see, for example, Corstange 2009; Trappmann et al. 2014 for a more detailed discussion). For a high level of statistical efficiency, it is desirable that the variance of the short list (non-key items) is small. To achieve this, it is preferable that the number of non-key items or “other persons” is low, that they have a prevalence near 0 or 1 (low variance), that they are negatively correlated with the sensitive item, and also negatively correlated among each other. Homophily among the uninvolved persons and the respondents themselves counteracts these ideals, because it causes high variance in the answers (people will tend to cluster at the minimum and maximum), and thus a large PCT standard error. In the basic ICT design, these features can be controlled by an appropriate and careful choice of the non-key items. For PCT, things are more difficult, because the researcher does not choose the “other persons” whom the respondents are asked to imagine. Hence, it is only the length of the short list (the number of uninvolved persons) that is directly controllable by design. As, for example, Wolter and Laier (2014, p. 155) recommend with respect to the ICT literature, a list length of three to five non-key items seems to be a good choice.

Another problem that could be more pronounced in the case of PCT than with ICT are what Blair and Imai (2012) call design effects. Both ICT and PCT rely on the assumption that respondents’ answers to the non-key items or the “other persons” do not change if the sensitive item or the respondent himself is added to the long-list group. If this happens, the mean difference of the short-list and long-list group is not exclusively determined by the sensitive item under concern (the addition of the respondent to the list in the PCT procedure), and the prevalence estimate is biased. When using PCT, the respondent’s own status for the sensitive item might, for example, affect his or her assessment of the status of the other persons in the list, causing design effects. This again would be an effect of (misperceived) homophily. With respect to this potential issue, further research including qualitative studies and cognitive pretesting should examine the likelihood of such design

effects. Pragmatically, Blair and Imai (2012) propose a statistical test empirically testing whether design effects have occurred.

One further constraint of PCT is that the so-called double list design cannot be implemented in a straightforward manner. Double list designs (Biemer et al. 2005; Droitcour et al. 1991) can improve the efficiency of ICT estimates considerably. The double list procedure administers a second short list of non-key questions to the respondents. Those in the original short-list group receive this second list including the (same) sensitive item; respondents in the former long-list group answer the second list without the sensitive item. The estimates from both lists can then be combined, resulting in lower standard errors than with only one list of innocuous questions. With PCT, this logic does not work because there is only one short list of "other persons". A remedy would be to introduce a second list of (different) people, but this seems to overcomplicate matters.

The issues discussed above reveal that the newly proposed PCT brings some challenges with it requiring further methodological and empirical research on how design aspects of PCT procedures affect the mechanisms at work and the statistical properties of the resulting estimates. This research should clarify whether the gain in simplicity of PCT vis-à-vis ICT outweighs the difficulties inherent to PCT and whether and how these problems can be resolved.

Besides these statistical aspects of ICT and PCT designs, the essential purpose and main goal of using these techniques remains achieving valid survey responses. With respect to ICT, a comprehensive meta-analysis of studies investigating the efficacy of ICT procedures with regard to avoiding or alleviating response bias is, to my knowledge, still lacking. Existing (summary) studies, however, do point, at least partially, to the result that ICT is successful in reducing response bias:¹ A small meta-analysis by Tourangeau and Yan (2007) of seven studies in which ICT was compared to DQ finds an overall positive, but non-significant ICT effect. A literature review by Wolter and Laier (2014) counts 22 comparative studies, of which 17 find results that are at least partially in favor of ICT. Two studies with aggregate external validation data in the field of voting behavior (and self-reporting on it) both find that ICT performs better than DQ with respect to response bias, but ICT estimates are still off the mark with regard to the externally validated true value (Comşa & Postelnicu 2013; Rosenfeld, Imai, & Shapiro 2015).

In terms of PCT, Grant et al. (2014) themselves provide a first empirical assessment of its performance as compared to DQ. In a telephone survey among registered voters in Illinois, respondents were asked about their intentions to vote for certain types of candidates in presidential elections. The PCT design corresponds to the one introduced above in this paper. The authors first find significant

1 However, it should also be noted that this does not mean ICT should be taken for granted as a universal remedy for all problems induced by sensitive questions. See, for example, Thomas, Johann, Kritzinger, Plescia, and Zeglovits (2017) for a critical study.

evidence for design effects regarding the Republican candidate item (which, for this reason, is not analyzed any further in the rest of the paper), and no evidence for design effects for the other four items. Second, PCT estimates of respondents claiming to be ready to vote for the respective type of presidential candidate are significantly lower than their DQ counterparts regarding the Democrat, female, and Mormon candidate (with a difference of about 20 percentage points). This is in line with the hypothesis that survey respondents, due to social desirability, claim to be open-minded and devoid of prejudice when asked directly, which results in overreporting in this case. For the latter item (“tea party member”), no difference is found between question modes.

Study Design and Methods

Survey Design

The PCT-DQ comparison for attitudes towards refugees was part of a local postal survey in the city of Mainz (Germany). The survey went by the title “Living and Residing in Mainz” and contained questions on a variety of topics: of the two main parts of the questionnaire, one was devoted to environmental problems, the other to attitudes and behaviors regarding foreigners and refugees/asylum seekers. Field work was carried out in autumn 2016. It should be noted with regard to the topic of refugees that within this period of 2015/2016, large numbers of asylum seekers, mainly from Syria and Afghanistan, came to Germany, which, in turn, created considerable concern and tension in the political debate and among parts of the German population.

Because one aim of the survey (not related to the topic of this paper) consisted in obtaining georeferenced data, we employed a special sampling design. Following an idea of Bauer (2014), we conducted a street section sample. Using GIS software for geographical data, we first identified all residential areas within the municipal area of Mainz and then randomly distributed 200 sampling points within these areas. For each of these (preliminary) sampling points, we then established the geographically nearest street sections, street section referring to the section between two street intersections (footways included). We then counted the number of households in each street section, yielding a number of 11,208 households. Another random sample of 68 street sections was then drawn from the original 200 sampling points, containing about 4,000 households.² Finally, every second household was

2 This procedure was necessary because the number of households in each street section was not known in advance. At the same time and for the purpose of other planned (multilevel) analyses, the number of cases in each sampling point had to be sufficiently high. Hence, we applied the two-step procedure of drawing 200 initial sampling points

manually assigned a questionnaire package. The package included a cover letter and a stamped envelope in order to send back the filled-out questionnaires without postage costs. We used the next-birthday method to randomly choose an adult person within each household. This sampling design leads to the selection probability decreasing for persons in larger households. However, I abstain from using design weights for the analyses, since the main goal of this study is the experimental comparison between DQ and PCT.

Out of 2,000 distributed questionnaires, 580 were returned, which corresponds to an AAPOR response rate of 29 percent (RR2). Because this study was a pilot study within the framework of a teaching project with MA students in sociology and, therefore, without funding, we were not able to dispatch follow-up letters or questionnaires to respondents who did not reply after the initial distribution of questionnaires.

The survey featured an experimental split into two subsamples. One half of the respondents were assigned to the PCT version of the questions on refugees, the other half to the DQ version. The DQ version also contained the short list of the PCT design. Normally, one would prefer to form three subgroups (DQ, short list, long list), but due to the financial restrictions of this study, we chose not to in order to ensure a sufficiently high number of cases in each group. However, this means that the samples yielding the DQ and PCT estimates are not independent from one another, which in turn requires special statistical procedures for the empirical analysis (see below).

In the analysis sample, 49 percent of all cases are in the DQ/PCT short-list group and 51 percent are in the PCT long-list group. This corresponds almost exactly to the 50-50 partitioning envisaged by the design. Table 1 reports the distribution of some socio-demographic variables by question format. There are no significant differences between the two experimental groups, meaning the randomization worked as intended. Women are slightly over-represented in the sample, as are people with higher education.

PCT Procedures

The PCT procedure was located roughly in the middle of the questionnaire within a block of various questions on attitudes, contact, and behaviors vis-à-vis refugees and immigrants in general. The PCT questions were devoted to aspects regarding refugees in the city of Mainz. The exact question wording for the long-list group

first, counting the households, and then drawing a subsample in order to meet the pre-defined distributional criteria by simultaneously not exceeding the projected sample size of 2000 contacts. Counting was carried out manually on location by sociology students.

Table 1 Distribution of Socio-Demographic Variables by Question Mode

	All	DQ	PCT	t	n
Gender (0 = male, 1 = female)	56.0	54.1	57.8	0.88	568
Age	49.6	50.1	49.1	0.56	564
Years of education	14.1	14.2	14.0	0.59	545
Social status (subj., [1...10])	6.3	6.3	6.3	0.36	570
House owner (0 = no, 1 = yes)	39.0	38.9	39.1	0.04	569
Married (0 = no, 1 = yes)	43.3	40.3	46.2	1.41	566

Note: DQ = direct questioning, PCT = person count technique. Reported are percent values (categorical variables) and means (metric variables). Differences between experimental groups were tested using t-tests (assuming equal variances).

and the four sensitive items are depicted in Figure 1 (translated from the German original).

There are three things to note on this design. First, the instruction asked respondents about “preferably diverse persons”. This was done in order to avoid homophily effects and, thus, to reduce the likelihood of floor and ceiling effects. Second, the design asked respondents to write down the initials of the first names of their imagined persons. On the one hand, pretests had shown that this helps respondents in coping with the questioning procedure. On the other hand, it is desirable that respondents do not switch around the people they are thinking of depending on the question content or the respondent’s own opinion (or for other reasons such as lack of knowledge about the persons of whom they initially thought). Of course, this is not a problem as long as the switching behavior is similar in both groups. However, the stimulus of including oneself in the long-list group might result in a different manner of switching and, hence, trigger design effects and biased results. By letting respondents write down the initials of their imagined persons we hoped to avoid this. Third, we did not introduce the PCT procedure as an “anonymizing technique” for “sensitive questions” or the like in order to avoid the respondents framing them in the sense of “the next questions are really sensitive”, which could be detrimental to the aim of achieving valid estimates. Also, this makes the questionnaire instruction more comparable to the short-list version of the PCT procedure. Furthermore, we anticipated that floor and ceiling effects could occur, resulting in a disclosure of the respondent’s individual answer. Introducing PCT as a technique that guarantees anonymity would represent a contradiction if this occurred and could lead to doubts or protests among respondents.

The following questions are about the situation in Mainz.

We are going to use a special questioning technique. For this purpose, please think of three preferably diverse persons among your friends, acquaintances or relatives who you know well and who live in Mainz, too. You can write down the initials of the first name of the three persons in the fields below – this makes things easier, but your notes will remain anonymous.

Initials of my three persons:

Now we are going to make a few statements for which you should estimate how many of these three persons plus yourself agree with the respective statement. The answer is thus a number between 0 (applies to no one) and 4 (applies to all three of the persons and yourself). If you are not sure, it is OK to guess, this is not a problem.

[Item 1] “I feel bothered by the refugees in Mainz”.

Number of persons who agree:

[Item 2] “Refugees should not stroll around in the city center of Mainz, but stay in their asylums”.

Number of persons who agree:

[Item 3] “I have a problem with refugees hanging out in my neighborhood”.

Number of persons who agree:

[Item 4] “The opening of a refugee asylum in my neighborhood would bother me”.

Number of persons who agree:

Note: Translated from the German original. Underlining is depicted as in the original.

Figure 1 Wording of the PCT Procedure (Long-list Group)

The wording in the short-list group was identical to the one presented in Figure 1, with the important difference that respondents were asked only about “three people” without themselves and to report a number between zero and three. As the short-list version of the questionnaire also contained the DQ questions of the four sensitive items, immediately after the short-list PCT procedure, the questionnaire read “And now we are interested in your personal opinion on these questions. Please answer with ‘yes’ or ‘no’”, followed by the same four items as in the PCT long-list version.

Methods

The survey design with only two (DQ and PCT short list versus PCT long list) instead of three experimental groups means that DQ and PCT estimates are not statistically independent from one another. This must be taken into account when calculating standard errors. Therefore, I calculated the mean estimates for DQ and for the short-list and long-list group, respectively, and used the Stata routine `suest` (seemingly unrelated estimation) in order to obtain a combined and robust covariance matrix. Tests for mode differences were then performed using this covariance matrix (cf. Weesie 1999).

As explained above, design effects are a potential problem of item count procedures. They occur if the addition of the sensitive item (or the respondent in PCT) to the long list affects the responses to the non-key items (“other persons” in PCT). I will follow the recommendations of Blair and Imai (2012) who propose a statistical test in order to empirically test for design effects. This test basically examines whether implausible negative proportions of respondent types (i.e., respondents with a certain combination of “yes” answers) arise if the sensitive item (the respondent himself or herself) is removed from the respective proportion of respondent type. If such negative proportions occur, the test calculates whether they could have arisen by chance. As the test’s logic and computation are complex, I refer to Blair and Imai (2012, pp. 63-65; see also Glynn 2013, pp. 165-167; Wolter & Laier 2014, p. 161) for further details. The test was performed using the “list” package for R by the same authors (Blair & Imai 2013).

Results

A conjecture made by some authors (e.g., Lensvelt-Mulders 2008, p. 464) is that sensitive questions result in higher item-nonresponse rates than non-sensitive questions. If this conjecture holds true and PCT works as intended, nonresponse should be lower when using PCT as compared to DQ. On the other hand, the PCT design requires more cognitive effort on the part of the respondents vis-à-vis answering a conventional survey question, which, in turn, could increase nonresponse rates. Table 2 shows the item-nonresponse rates for each of the four sensitive items regarding attitudes towards refugees in DQ mode and in the two groups of PCT mode.

In DQ mode, nonresponse rates for the four items vary from 2.1 to 2.8 percent, which can be considered low values given that this was a classic self-administered postal survey. This confirms the aforementioned position of Tourangeau and Yan (2007, p. 862) that item-nonresponse generally does not pose a serious problem for sensitive questions. Nonresponse rates for the PCT long-list group are higher and amount to roughly 6 percent. The differences with respect to DQ are all significant

Table 2 Item-Nonresponse Rates by Question Mode

		Item 1	Item 2	Item 3	Item 4
DQ (n = 284)	% NR	2.46	2.82	2.11	2.11
PCT LL (n = 296)	% NR	5.74	6.08	6.08	6.08
PCT SL (n = 284)	% NR	7.04	6.69	6.69	7.04
χ^2 DQ-PCT LL		3.93 *	3.61 +	5.75 *	5.75 *
χ^2 DQ-PCT SL		11.27 ***	8.07 **	11.27 ***	12.25 ***
χ^2 PCT LL-PCT SL		0.41	0.09	0.09	0.22

Note: DQ = direct questioning, PCT = person count technique, LL = long list, SL = short list, NR = nonresponse. Differences were tested using conventional χ^2 tests for differences between experimental modes and McNemar's χ^2 statistic for the DQ-PCT short-list difference. + $p < 0.1$; * $p < 0.05$; ** $p < 0.01$; *** $p < 0.001$.

at least on a 10 percent level. However, the higher nonresponse rates do not seem to be attributable to PCT causing the items to appear more sensitive to the respondents (which in turn could yield higher nonresponse rates), because the nonresponse rates for the PCT short-list version are similar to those from the long-list group and not significantly different from them. Instead, it is the PCT design per se – be it the short or the long list – which boosts nonresponse rates, presumably due to its cognitive demands. Of course, this would be a drawback of this new questioning technique. However, it should again be noted that the survey was self-administered with no interviewer present. Taking this into consideration, nonresponse rates of 6 to 7 percent do not appear to be exceedingly or unreasonably high. Further studies should examine to what degree interviewer-administered survey modes can provide a better approach in order to avoid item-nonresponse in PCT designs.

Before we look at the prevalence estimates for the four sensitive items depending on question mode, Table 3 reports information on the distribution of respondents' answers in the short-list and long-list group, respectively. What is important here are floor and ceiling effects, i.e., respondents denying or affirming all items or persons in the list. Above I made the case for the assumption that, when using PCT instead of ICT, floor and ceiling effects will be more problematic because of homophily effects.

The results in Table 3 clearly confirm this assumption. Floor effects are substantial for all four items, both for the short-list and the long-list groups. Up to 77 percent of respondents report that the sensitive item applies to none of the persons of whom they had been asked to think. For the long-list group, containing the respondent himself or herself, anonymity is no longer ensured. However, given that a “yes” answer to the sensitive item corresponds to expressing a socially undesir-

Table 3 Distribution of Answers in the Short-List and Long-List Group

	Item 1 (%)		Item 2 (%)		Item 3 (%)		Item 4 (%)	
	<i>SL</i>	<i>LL</i>	<i>SL</i>	<i>LL</i>	<i>SL</i>	<i>LL</i>	<i>SL</i>	<i>LL</i>
0	56.8	49.5	77.0	73.7	44.5	35.6	27.7	21.6
1	22.7	23.7	16.2	13.0	30.9	30.9	31.1	19.8
2	14.4	11.5	4.2	7.2	16.6	18.4	23.1	25.2
3	6.1	8.6	2.6	2.9	7.9	8.6	18.2	18.4
4	-	6.8	-	3.2	-	6.5	-	15.1
<i>n</i>	264	279	265	278	265	278	264	278

Note: *LL* = long list, *SL* = short list.

able attitude, these floor effects are probably less problematic regarding response bias. In this regard, ceiling effects, i.e., respondents reporting “4” for the long list are the main problem, because their sensitive answer is no longer concealed by the PCT design. This holds for approximately 7 (item 1 and 3), 3 (item 2), and 15 (item 4) percent of respondents. While 3 percent (corresponding to 9 out of 278 respondents) appear to be within an acceptable range, 15 percent for item 4 (42 out of 278 respondents) is definitely too high and endangers the main purpose of PCT, namely assuring anonymity. At first glance, this appears to be a major drawback of PCT as compared to the classic ICT design, wherein floor and ceiling effects can be prevented by a careful design of the non-key items. Further studies should investigate possibilities to avoid floor and ceiling effects in PCT designs. For the time being, I suggest following our PCT design reported in Figure 1 above and, at least for now, to not all too loudly hail PCT as a technique that “guarantees complete anonymity”. Future research should also investigate whether the wording of the items affects the tendency for floor and ceiling effects. For example, for item 2 (Table 3), the fraction of “0” answers is by far the highest among the four items. In addition to substantive reasons regarding the level of sensitivity of this item, it can be assumed that this is due to the different cognitive demands processing a single sentence (item 1, 3, and 4) vis-à-vis a normative statement (item 2) requires.

Besides looking at floor and ceiling effects, I performed the aforementioned test for design effects as proposed by Blair and Imai (2012). For none of the four sensitive items could I find evidence for such effects, the *p*-values for items 1 to 4 are, respectively, $p = 0.72$, $p = 0.69$, $p = 1.00$, and $p = 1.00$ (the null hypothesis is that there are no design effects; thus, the null cannot be rejected according to the *p*-values). This can be interpreted as being in favor of PCT, because, at least empirically, based on the Blair-Imai test, there is no evidence that including the respon-

dents themselves in the PCT-long-list changes response behavior to the “other persons” in the list.

Table 4 reports the main results of the study, namely the prevalence estimates of the four sensitive items on attitudes towards refugees in Mainz, according to question formats DQ and PCT. As expressing hostile attitudes towards refugees is considered socially undesirable, higher estimates are taken as more valid than lower ones. Therefore, the DQ-PCT comparison is based on the “more is better” assumption.

The estimates of dismissive attitudes towards refugees are substantially higher in PCT mode than in DQ mode. This holds for all four items. Regarding item 1 and 2 (“I feel bothered by the refugees in Mainz”; “Refugees should not stroll around in the city center of Mainz, but stay in their asylums”), the PCT estimates are three times higher than the DQ ones. However, as the *z* statistics show, PCT-DQ differences are statistically significant for the first item only, while for item 2 to 4, DQ estimates are not significantly different from their PCT counterparts at conventional levels. An overall test for the DQ-PCT difference, taking into account the four items simultaneously and adjusting for the clustering by respondents, also fails to reach conventional significance levels ($\text{diff} = 12.21$, $z = 1.46$, $p = 0.145$). These results are due to the highly inflated standard errors of the PCT estimates. For example, the estimate of 54 percent “yes” answers for item 4 comes with a standard error of more than ten percentage points. As pointed out above, standard errors of ICT estimates will always be higher than those from conventional ones. However, the PCT procedure, as it was implemented in this study, probably aggravates this issue for several reasons. On the one hand and as shown above, there are many respondents who answer “zero” to the person list, and a non-negligible fraction states that the trait applies to all persons in the list. This pattern inflates the variance of the variables, which, in turn, leads to greater standard errors. In classic ICT with non-key items that have either a high or low prevalence, the variance will usually be lower and the standard errors will also follow suit. On the other hand, the correlation between the sensitive item (in PCT: the respondent himself) and the filler items (in PCT: the “other persons”) is probably not negative due to homophily effects, which also boosts standard errors. Furthermore, the prevalence of the sensitive trait itself will also have an impact on standard errors, because the variance of binary variables is a function of their mean and highest for an equal distribution (i.e., a prevalence of 50 percent). These considerations show that careful precautions are required when developing PCT designs. Further studies should go into more depth on these issues and examine the relationship between design features and statistical properties of PCT estimates in a more general perspective.

Despite these challenges and despite the lacking significance for three out of the four items examined in this study, the overall conclusion remains in favor of PCT with respect to its potential and the validity of its estimates: For all items, the

Table 4 Prevalence Estimates of the Sensitive Items by Question Format

		Item 1	Item 2	Item 3	Item 4
DQ	% "yes"	9.75	5.43	23.74	43.88
	s.e.	1.78	1.37	2.55	2.98
	n	277	276	278	278
PCT	% "yes"	29.94	16.47	31.50	53.79
	s.e.	9.43	7.20	9.28	10.44
	n (short list)	264	265	265	264
	n (long list)	279	278	278	278
Difference		20.20	10.72	7.76	9.91
z		2.00 *	1.43	0.75	0.83

Note: DQ = direct questioning, PCT = person count technique. Standard errors and test statistics were calculated taking into account that DQ and PCT estimates are not independent from one another (see the "Methods" section for details). * $p < 0.05$.

direction of the DQ-PCT comparison points in the anticipated direction. Most of the respondents were able to cope with the PCT instructions without assistance of an interviewer and nonresponse rates were not unreasonably high.

Discussion

The present study evaluated the performance of PCT in a mode-comparing perspective and investigated item-nonresponse and underreporting on four questions regarding hostile attitudes towards refugees in a German city. As far as nonresponse is concerned, the observed rates are higher in PCT mode than in DQ mode. This, however, seems not to be caused by the sensitivity of the questions being asked, but by the PCT procedure itself, which was implemented here in a self-administered postal survey. Despite being higher, nonresponse rates remain at a tolerable level also in PCT mode. With respect to the prevalence of the four sensitive items, all estimates are distinctively higher in PCT mode, though significantly different from DQ for one item only. In this context, very large standard errors of the PCT estimates have been observed, presumably caused by the distribution of answers regarding the "other persons" in the item lists and their correlations among each other and with the respondent himself in the long list. All in all, however, the results show that considerable underreporting of hostile attitudes to refugees

occurs when using conventional questioning techniques. Although the findings are not unequivocally in favor of PCT, they suggest considering PCT as a promising alternative in future studies.

Aside from the general difficulties of PCT, this study has some obvious shortcomings, that should be taken into account when judging the results. First, the number of cases ($N=580$) was low. As the elevated standard errors of the PCT estimates show, a larger sample would have been much more preferable and should be aimed for in future studies. Because of the limited sample size, a two-group design (DQ and PCT short list versus PCT long list) had to be used instead of a three-group design with a random split into DQ-, short-list, and long-list subsamples. This two-group design means, firstly, that DQ and PCT estimates are not statistically independent, which has to be taken into account when performing tests for differences. Secondly, halo effects may affect the results because the experimental stimulus (PCT versus DQ) is confounded with question order. The limited statistical power was also the reason why I restricted the analysis to prevalence estimates and did not conduct a regression analysis on determinants of xenophobic attitudes. Such analysis could have been helpful in judging the external validity of the PCT estimates. Whereas regression analysis is generally possible with ICT (or PCT) data (Blair & Imai, 2012, 2013; Imai 2011), it requires large sample sizes due to the restricted statistical power of PCT data. Further, the elevated item-nonresponse rates of the PCT questions show that self-administered survey modes may not be the best choice when planning to use PCT procedures. Interviewer-administered surveys seem to be preferable in this regard. Another flaw is that validation of the PCT results could only be carried out here on the basis of a “more is better” assumption. As no true values were at hand, higher estimates of hostile attitudes to foreigners were assumed to be more valid. To what degree higher estimates are still off the mark from the true value remains undiscoverable with this approach.

Above, several challenges of PCT have been pointed out, namely floor and ceiling effects, statistical power issues, and design effects. In contrast to the classic ICT design, the researcher has less influence on addressing these issues via a thoughtful design of the non-key items. In what follows, I will propose some modifications or alternatives to the PCT design as it was implemented in the present study, which could (partly) address these issues.

A first modification of the original PCT design aims to give the researcher control over the characteristics of the “other persons”. This would help in avoiding floor- and ceiling effects and in making PCT estimates more efficient. I call this design fixed person count technique (FPCT). The simple idea is not to ask respondents to imagine “some people they know”, but instead to propose fixed persons by design. A (purely illustrative) example would be to ask respondents to indicate how many of the following persons, including themselves, have already smoked marijuana: Bob Marley, Angela Merkel, and Pope Francis. In this case, the values for

Bob Marley and Pope Francis are more or less fixed and near 1 and 0, respectively. This avoids floor and ceiling effects and improves statistical efficiency. For the sake of anonymity, the Angela Merkel item is more ambiguous. Of course, this is just an illustrative example, as one should not choose such obvious cases as Bob Marley and marihuana consumption. One could easily imagine other possible designs in this regard, for instance, letting respondents imagine a member of a typical group such as a “typical democrat voter” or a “typical primary school teacher”. Or, to think of their nearest neighbor, their postman, or their family doctor. A clever choice of these more or less fixed persons might help overcome the problems inherent to the basic PCT design.

Another straightforward modification of PCT is to apply the logic of the above-mentioned item sum technique (IST) for metric sensitive variables to a PCT procedure – the person sum technique (PST) as proposed by Junkermann (2018). PST also asks respondents to imagine one or more other persons they know – as with IST, however, one non-key person will usually suffice. Respondents are then asked to estimate the value of one quantitative sensitive item for the other person in the short-list group. In the long-list group, respondents are asked to add up the value of the other person and their own value. For example, the sensitive item could be the number of cigarettes smoked per day. Respondents in the long-list group are then asked to estimate the number of daily smoked cigarettes for the uninvolved person, and to add this value to the number of cigarettes smoked by themselves.

The research desiderata with respect to PCT are clear-cut. Future studies should, firstly, investigate the (cognitive) mechanisms at work when respondents deal with PCT designs. These studies should focus on, among other matters, homophily effects, isolated persons that have difficulty imagining people they know well, the occurrence of design effects, and what happens if respondents are unsure about the status of the uninvolved person(s) in the list. This entails both qualitative and quantitative work. Second, real validation studies with known true values (from external records, for instance) should be conducted in order to assess the ability of PCT to avoid or at least alleviate response bias. If this is not possible, further studies relying on the “more is better” logic should be conducted – and with larger samples than in the study presented in this paper. Third, empirical studies should also concentrate on experimentally comparing PCT with classic ICT designs. This should be carried out with respect to validity, the anonymity protection subjectively perceived by the respondents, the amount of cognitive burden (is PCT really less demanding than ICT?), and with respect to the trade-off between statistical efficiency, respondent protection, and simplicity of the question procedures. Fourth, further studies on PCT designs should test whether the above introduced FPCT presents a viable alternative to the original PCT design.

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