Manuscript Title: Surveying the Surveyors to Address Risk Perception and Adaptive Behaviour Cross-study Comparability

Summary: This manuscript describes a uniquely intriguing and informative examination of research practices in the field of risk perception and protective action. The authors have addressed some important issues that are related to, but distinctly different from, those associated with the replication crisis (e.g., Shrout & Rodgers, 2018). The 150 member sample is sufficiently large to draw some reasonable conclusions even though it has a very high proportion of European researchers studying floods. In the absence of a census of researchers and hazards in the field of risk perception and adaptive behavior, it is unknown whether these sample characteristics are representative of the field’s researchers. Nonetheless, the authors have identified and summarized problems that I have seen repeatedly during my 50 years studying floods, volcanic eruptions, hazardous materials releases, earthquakes, hurricanes, tornadoes, and tsunamis in the US and other Pacific Basin countries. Consequently, their data are likely to generalize to other hazards and countries. In particular, their data on theoretical frameworks, hazards studied, research designs, and sources of explanatory variables for risk perception and protective action provide ample support for their six recommendations for improving comparability of research results. I strongly endorse publishing this manuscript, but also have some suggestions for improving it.

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<td>52</td>
<td>It is important to distinguish between inconsistent findings and contradictory findings. Inconsistent findings occur when a variable that has statistically significant correlation/regression coefficients in some studies but nonsignificant coefficients in other studies, whereas contradictory findings occur, for example, when a variable has coefficients of opposite signs in different studies. These problems can be addressed by statistical meta analyses showing that the average effect size is not significantly different from zero (for inconsistent findings) or that a moderator variable can account for the differences in the direction of effects (for contradictory findings). Of course, concluding that research findings are inconsistent does not explain why those findings are inconsistent but, as noted below, inconsistent findings about psychological variables are likely due to inconsistent operationalizations (i.e., indicator variables) of latent variables.</td>
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<td>62</td>
<td>There are other reasons for encouraging individual households to become more resilient, such as the large scope of destruction in a disaster preventing community emergency managers from providing immediate assistance. For example, emergency managers in some US Pacific Coast jurisdictions advise households that they must be self-sufficient for up to 14 days after a major earthquake because destruction of transportation infrastructure will prevent the authorities from delivering assistance.</td>
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<td>73</td>
<td>It would be helpful to be more specific about what is meant by “opposite conclusions”. Does this mean A does vs. doesn’t have a significant relationship with B, A has a positive vs. negative relationship with B, or A causes B vs. B causes A?</td>
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<td>87</td>
<td>The term “minimal requirements” could benefit from greater specificity. Does this refer to theoretical grounding, population sampling, psychological measurement (i.e., the reliability and construct validity of the latent variables), statistical analysis (e.g., compliance with statistical model assumptions), or other aspects of the study design?</td>
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| 184  | I was initially surprised that Response was only fifth on the list, but I noticed that Adaptive behavior was listed seventh, and Intention/behavior was listed eighth, so it is clear that behavioral issues are more important than they might seem from cursory examination of Figure 4. Further examination of this figure showed that other topics could also be grouped—for example, Worry, Fear, Denial, Fatalism, and Helplessness as affective responses. However, that categorization
would depend on some important theoretical distinctions, such as automatic vs. controlled processes (e.g., Moors & De Houwer, 2007), where Worry, Fear, and Helplessness might be considered examples of automatic processes, but Denial and Fatalism might be considered examples of controlled processes. It would be interesting to see these elements grouped according to their hierarchical clustering.

As with Figure 4, it seemed surprising initially that response/coping was only nominated once as the first item and is third on the list of summed nominations. However, if Response/coping is combined with Adaptive behavior, the combined category would be first on the list of specific summed nominations with 53 mentions.

Figure 7 would be easier to understand if Comparison with other studies and Literature review, which seem synonymous, were adjacent. Similarly, Experience and habits and Previous (own) studies seem quite similar, so placing them adjacent to each other would allow the reader to better understand their total effect. Of course, if each pair of categories is exactly synonymous, then summing them is double counting. Nonetheless, it seems clear that reviewing the published literature has the greatest effect, followed by continuity of the researcher’s previous work, peers’ recommendations, and finally, exploratory innovations.

I am unsure how to interpret the statement that “only a minority considered ‘comparison with other studies’ relevant for their design choices”, given that the overwhelming majority (87%) reported relying on literature reviews.

The use of the term “panic” here might seem like a relatively innocuous example of the hyperbolic use of that term, but I find it concerning because it tends to perpetuate a myth that disaster researchers have long debunked (Fritz & Marks, 1954). The problem with using the label “panic” as anything other than a clinical diagnosis by a qualified mental health professional is its misuse by the news media and misinterpretation by public authorities (Clarke & Chess, 2008). In the present manuscript, the phrase “panic seems unnecessary” could be replaced quite satisfactorily by “there is no need for undue concern”.

Given the data in Figures 6 and 7, which show that research designs are heavily influenced by comparison with other studies and literature reviews, it seems inappropriate to characterize risk perception research as an “uncoordinated, exploratory” effort. A more appropriate term might be an “organized anarchy” (Cohen et al., 1972) of very loosely coordinated research efforts. I have found that the most experienced researchers (more or less defined by those with the most publications in the field), and less experienced researchers who have studied with the most experienced researchers, operate within self-defined domains that are coordinated implicitly (by recognizing the limits of their own research paradigms) and substantially confirmatory (assessing the generalizability of their paradigms by applying it to different hazards and populations). Instead, “uncoordinated, exploratory” research appears to be conducted mostly by the other tail of the experience distribution—people who are drawn into the field by a highly publicized event and conduct a survey that is designed to test their (often incorrect) intuitions about the relevant variables without reviewing the existing literature. Many such researchers publish a single study in a journal whose reviewers are equally unfamiliar with the hazards/disasters literature and then move on to another topic. Their studies’ weak designs and idiosyncratically measured constructs make it difficult to assess their contribution to the literature; they’re not bad enough to dismiss entirely but not good enough to make a solid contribution.

The prevalence of age, gender, education, income, family or household composition, and occupation as explanatory variables seems relatively easy to understand. Despite evidence of their
weak and inconsistent correlations with behavioral variables such as evacuation (Baker, 1991; Huang et al., 2016), it is conventional to measure these variables so readers can assess the degree to which a survey sample is biased in comparison to census data for the geographical area in which the data were collected. Since these demographic variables have already been collected, it is a simple matter to examine their correlations with psychological and behavioral variables. Other possible reasons for the prevalence of these demographic variables are that their inclusion requires no knowledge of the research area and that they are compatible with any of the theoretical perspectives listed in Figure 2 (which are not necessarily compatible with each other).

The reversal of the demographic, environmental, and contextual variables between frequency of use and relevance to cross-study comparison is extremely intriguing. One possible explanation is an extension of my previous comment about the prevalence of demographic variables. Specifically, the frequency of use data can be explained by the fact that it takes no imagination to propose demographic variables as predictors of protective action and not much more imagination to propose environmental variables (exposure and experience) as antecedents of risk perception. However, the significance of contextual variables (vulnerability and resilience) has only been recognized more recently. The different rank ordering with respect to relevance could be explained by the fact that environmental variables such as hazard experience tend to be homogeneous within communities, especially for infrequent hazards, so studies focused upon a single community or small region will find little variation in hazard experience. Thus, it is necessary to survey communities that differ substantially in their hazard experience to obtain the requisite variation in experience at the household level (e.g., the Lindell & Prater, 2000, comparison of low-seismic Western Washington cities with high-seismic California cities). Since such designs are much more expensive than single community designs even though they are much more informative, that could explain why the difference between use and relevance.

The correlation matrix in Figure 9 is very useful, but it could be improved by conducting a hierarchical cluster analysis and then rearranging the variables according to the resulting clusters. For example, the current figure clearly indicates that age, gender and education form a cluster but closer examination suggests that other variables also belong in this cluster—definitely income and probably the housing variables.

The presentation of the results in Figure 11 would benefit from an explanation of the questions that were asked. For the upper left panel, were respondents asked in an open-ended question to identify the hazards they studied and then the responses were categorized as presented in the upper left panel? Were the data in the upper right panel generated by asking respondents to report the total number of communities they had studied or the maximum number of communities in the study with the largest number of communities? Were the data in the lower right panel generated by asking respondents to report the total number of respondents in all of their risk perception studies, the typical number in any of their risk perception studies, or the number of respondents in the study with the largest number of respondents?

It would be helpful to clarify the comparison group in the statement “While respondents exclusively using interviews or focus groups were more likely [than survey researchers?] to use contextual factors”.

I don’t understand the rationale for the statement “Although the sample size is not appropriate to use inferential statistics due to high data scattering”. First, it appears from Figure 12 that there are enough cases to test the difference between European and non-European researchers in their use of age and many of the other variables at the top of the variable list. Of course, at the other extreme, it seems quite clear that there definitely are not enough cases to test such differences in
the use of language and many of the other variables at the bottom of the variable list. Similarly, it appears that there are enough cases to test the difference between studies in European, Mixed, and non-European locations on some, but not all, of the explanatory variables. Thus, there appears to be no basis for neglecting all statistical comparisons because of sample size.

Second, the authors seem to be using “high data scattering” to mean “large variance”, but it is unclear if this means variance in counts across explanatory variables, across researcher locations/study locations, or both. Whatever the authors’ intended meaning, it is unclear why this variation would be a problem. As noted in the paragraph above, there appears to be an adequate sample size for assessing differences on some, but not all, of the explanatory variables.

340 The statement “However, this observation is blurred … indirectly informed by specific risk and vulnerability theories” seems quite plausible, but the authors should provide the data analyses that support it or clearly acknowledge that it is a speculative explanation. The following statement “Another regional discrepancy…inside their country” does not have this problem because it is explicitly stated as a speculative explanation—“may be explicable from a historical perspective” (my emphasis).

377 It is true that an effect of flood experience is commonly explained by the availability heuristic, but flood experience was used as an explanatory variable for risk perception and hazard adjustment much earlier than Tversky and Kahneman (1974) proposed the availability heuristic (e.g., Kates, 1963). More generally, Wyer & Albarracin (2005) expanded on the concept of availability by noting that the four factors influencing the retrieval of belief-relevant knowledge are recency, frequency, strength of association, and schematic processing. More recently, Demuth (2018) proposed a comprehensive method of measuring experience.

404 Research hypotheses (e.g., Var1 is positively related to Var2) are one important implication of a theory, but research questions (Is Var1 significantly related to Var2?) can also make a significant contribution to the research literature. Research questions are especially relevant when there are conflicting results in the research literature (e.g., a mixture of significant positive, nonsignificant, and possibly even significant negative relations) that preclude a research hypothesis.

406 Just as important as the derivation of constructs from a theoretical framework is the operationalization of those constructs in terms of indicators (e.g., questionnaire items) that are either consistent with previous operationalizations or are derived from a different theoretical orientation. For example, American disaster researchers define risk perception in terms of personal consequences (e.g., Mileti et al., 1992) that are distinctly different from the risk dimensions in Slovic’s (1987) framework. As suggested above, inconsistency in research findings can often be attributed to differences in the operationalization of constructs such as experience and risk perception (Lindell & Perry, 2000).

409 One nonobvious implication of Question 6 is the need for researchers to report the interrelationships among all of the variables that have been measured to test their research hypotheses. Unfortunately, some researchers report only the variables that have significant correlation/regression coefficients with a study’s dependent variable(s) such as risk perception and behavior/behavioral intentions. This is a variant of the “file drawer problem” (Rosenthal, 1979). This mistaken practice distorts the scientific record by inflating the estimated effect sizes for the reported variables in statistical meta analyses because only larger, statistically significant, effect sizes are reported in the literature.

425 One limitation to the recommendation for emphasizing the “currently most frequently used
questions and explanatory variables” is that Figure 11 shows that these are demographic variables, which have been found to have small and inconsistent effects, as least in the case of hurricane evacuation (Baker, 1991; Huang et al., 2016). Moreover, this conclusion is consistent with Fishbein and Ajzen’s (1975) proposal that demographic variables affect behavior indirectly through their effects on psychological variables. Conversely, the attributes of protective actions do not appear in Figure 11’s variable list at all despite Fishbein and Ajzen’s proposal that an attitude toward an object (i.e., risk perception) is less predictive of behavior than attitudes toward actions related to that object (protective actions/adaptive behaviors). In PMT, the attributes of the protective actions are response-efficacy and response costs (e.g., Floyd et al., 2000). In the PADM (Lindell, 2018; Lindell & Perry, 2012), they are hazard-related attributes (protection of persons and property, utility for other purposes) and resource-related attributes (cost, and requirements for knowledge/skill, time/effort, tools/equipment, and social cooperation).

The issue of regional differences can be addressed very effectively in statistical meta analyses.

The term “temporal and analysis scale variation” is unclear and needs to be defined.

References