

1 **Investigating the suitability of dichotomous responses for the Water Insecurity**
2 **Experience (WISE) Scales using nationally representative data from 39 countries**

3

4 **Authors:** Indira Bose^{1,2}, Joshua Miller³, Hilary J. Bethancourt^{4,5}, Olga P. García⁶, Hugo
5 Melgar-Quiñonez⁷, Rosana Salles-Costa⁸, Mauro E. DelGrossi⁹, Scott Miller¹⁰, Tessa
6 Durham¹⁰, Teresa Shamah-Levy¹¹, Pablo Gaitán-Rossi¹², Sera L. Young^{5, 13} and Edward A.
7 Frongillo¹⁴

8

9 **Affiliations:**

10 1. Department of Population Health, London School of Hygiene & Tropical Medicine, U.K.

11 2. The Centre on Climate Change and Planetary Health, London School of Hygiene &
12 Tropical Medicine, U.K.

13 3. Department of Nutrition, University of North Carolina at Chapel Hill, USA.

14 4. Department of Medicine, University of California, San Francisco, USA.

15 5. Department of Anthropology, Northwestern University, USA.

16 6. School of Natural Sciences, Universidad Autónoma de Querétaro, México.

17 7. School of Human Nutrition, McGill University, Canada.

18 8. Institute of Nutrition Josué de Castro, Federal University of Rio de Janeiro, Brazil.

19 9. University of Brasilia, Brazil.

20 10. charity: water, USA.

21 11. Centro de Investigación en Evaluación y Encuestas (CIEE), Instituto Nacional de Salud
22 Pública (INSP). Cuernavaca, Morelos, México.

23 12. Research Institute for Equitable Development (EQUIDE), Universidad Iberoamericana,
24 Mexico City, Mexico.

25 13. Institute for Policy Research, Northwestern University, USA.

26 14. Arnold School of Public Health, University of South Carolina, USA.

27

28 **Abstract**

29 **Background**

30 The Water Insecurity Experiences (WISE) Scales have been validated to comparably
31 measure water insecurity globally. The scales consist of 12 items that can be administered in
32 approximately 3 minutes. There is interest in developing more rapid WISE Scale versions,
33 for use when time is limited. One alternative is to use a subset of 4 items, which has been
34 validated, but has some drawbacks. Here we investigate another alternative: dichotomous
35 (yes/no) response options instead of the original four levels of frequency-based (polytomous)
36 responses.

37

38 **Methods/principle findings.**

39 We used nationally representative data from 39 countries to simulate dichotomized
40 responses by collapsing the four levels of frequency (never, rarely, sometimes, often/always)
41 into yes/no. We first explored if “rarely” is meaningful in the gradation of water insecurity, as
42 experiences that occur “rarely” may not be affirmed with dichotomous response options. We
43 tested item-by-item if “rarely” responses predicted dissatisfaction with water quality using
44 logistic regression and found that they were associated with higher odds of dissatisfaction
45 with water quality. As such, some meaningful nuance may be lost if “rare” experiences are
46 not affirmed as “yes”.

47

48 We then compared the predictive accuracy of WISE scores using simulated dichotomous
49 responses vs. those calculated using polytomous responses. Based on ROC curves and
50 regression models, dichotomized response scores had good predictive accuracy. Scores
51 calculated using the abbreviated 4-item version were also found to be similarly accurate.
52 Finally, we explored if it was possible to determine levels of water insecurity that were
53 comparable to the levels calculated using the original responses using dichotomized
54 responses. Using ROC curves, we found that it was possible, which is an advantage over
55 the 4-item scales.

56

57 **Conclusion/significance**

58 Although polytomous responses options provide more information, dichotomous response

59 options hold promise as a quicker alternative for measuring water insecurity.

60 Introduction

61 Water insecurity, the inability to reliably access sufficient water for basic domestic needs, is
62 of increasing global concern given its negative effects on health and well-being. Accurate
63 and reliable measurement of water insecurity is critical for identifying affected and vulnerable
64 populations, developing policies and programs to reduce its burden, and advancing progress
65 toward Sustainable Development Goal 6, ensuring access to safe water for all (1-4).

66

67 The Water Insecurity Experiences (WISE) Scales comparably measure experiences with
68 issues with reliable access and use of water for basic domestic needs across countries (5).
69 The Scales comprise 12 questions that can be used for assessments at the household (the
70 HWISE Scale) (3) and individual level (the IWISE Scale) (6); they take approximately 3
71 minutes to administer (5). The WISE scales complement existing “provider-side” water
72 indicators (e.g., access to safely managed drinking water services using WHO/UNICEF’s
73 Joint Monitoring Programme’s criteria (7)) by capturing the state of people’s water insecurity
74 (8). That is, they offer “user-side” perspectives on experiences accessing and using water for
75 consumption, hygiene, and other basic activities. In recognition of the added value of these
76 scales to existing measures, they are increasingly being adopted by governments and
77 development organizations to inform policy and practice and to monitor and evaluate the
78 impact of water-related programs (5, 9-11).

79

80 In the original versions of the HWISE and IWISE Scales, item responses measured not only
81 if each of the 12 water-related issues was experienced, but also the frequency (“never”,
82 “rarely”, “sometimes”, or “often/always”) with which they were experienced in a specified
83 recall period (**Figure 1**). For example, “rarely” refers to an experience occurring on 1-2 days
84 over a 4 week recall period or in 1-2 months over a 12-month recall period (**Figure 1**). Each
85 response is scored 0-3; the sum of the item responses can then be used to classify
86 individuals or households as experiencing no-to-marginal, low, moderate, or high water
87 insecurity (12).

88

89 **Figure 1.** *The items in the Water Insecurity Experiences (WISE) Scales ask about adverse*
90 *experiences caused by problems with water, represented by these 12 icons and labels.*

91 *Twelve items take approximately 3 minutes to administer. The four items on the top row are*
92 *those that comprise the abbreviated versions and take one minute to administer. Full*
93 *phrasing is in Supplemental Table 1. Figure reproduced from a prior publication (13).*

94 Note: Polytomous response options are “Never”, “Rarely”, “Sometimes”, “Often/Always”; dichotomized response
95 options are “No” or “Yes”.

96

97 Academics, as well as those in governmental and non-governmental organizations working
98 with constrained budgets or rapid assessments, have expressed interest in developing
99 versions of WISE scales that are quicker to implement and reduce participant burden. To
100 that end, abbreviated versions of the WISE scales – the HWISE-4 (14) and IWISE-4 (13)
101 Scales – which are composed of a subset of four items and take only one minute to
102 administer, have been developed and validated in low- and middle-income countries (13, 14)
103 (**Figure 1**). In March 2024, the IWISE-4 Scale was recommended by WHO/UNICEF’s Joint
104 Monitoring Programme for generating gender-disaggregated indicators to monitor global
105 progress toward Sustainable Development Goal 6 (15).

106

107 Although the 4-item WISE Scales have practical benefits, they are limited in two important
108 ways. First, because they only query four experiences, they may not capture key
109 experiences of water insecurity in some settings. For instance, water interruptions were the
110 most commonly reported item by respondents to the 2022 Gallup World Poll in Australia and
111 the United States (16), but “interruptions” is not an item included in the abbreviated scales.
112 Second, these abbreviated scales do not generate enough information to assess multiple
113 categories of water insecurity. That is, the 4-item WISE Scales can only categorize
114 individuals or households as experiencing water insecurity, or not, and are not well-suited to
115 measure different severity levels along the spectrum. Given the global heterogeneity in

116 experiences of water insecurity and the effects that even low levels of water insecurity can
117 have on health and well-being, the ability to classify individuals into multiple categories of
118 water insecurity using the 12 WISE items has practical utility (12).

119

120 ***Dichotomous Response Options: An Alternative for Rapid Assessment***

121 Given the limitations of the 4-item WISE Scales, there is interest in investigating other
122 strategies for measuring water insecurity experiences efficiently while maintaining validity for
123 situations in which rapid data collection and limited resources are key concerns. One such
124 strategy is using dichotomous response options, whereby respondents report whether they
125 ever experienced any of the issues (i.e., “yes” or “no”) within a particular recall period.

126 Dichotomous responses with experience-based food-insecurity scales have been used at
127 the national, regional, and global level for decades (17-19). In Brazil, for example, the
128 Brazilian Scale of Food Insecurity has been useful for identifying categories of food
129 insecurity, enabling policymakers and program designers to address the distinct causes and
130 solutions associated with each severity category (20).

131

132 Scales with dichotomous response options offer several potential advantages over those
133 with polytomous levels (e.g., never/rarely/sometimes/often). They may be quicker to
134 administer and therefore less costly, making them more feasible when budget and time are
135 limited. For example, a study in Mexico found that prompting with dichotomous response
136 options on the HWISE scale reduced the time to administer the scale from three to two
137 minutes (21). The reason for this may be that dichotomous response options reduce the
138 cognitive burden on respondents and enumerators; respondents are not required to recall
139 the exact frequency of each experience and prompting by enumerators can be reduced.

140 Furthermore, dichotomous responses can be more stable than polytomous responses when
141 assessing a scale’s psychometric properties and equating it across countries using Rasch
142 models (22), the current modeling approach used with the Food Insecurity Experience Scale
143 (FIES) across countries worldwide to track progress toward Sustainable Development Goal

144 target 2.1 (23, 24). For instance, if responses of “rarely” are less frequent than responses of
145 “sometimes”, this can lead to disordered thresholds (i.e., severity levels that do not increase
146 monotonically across response categories) in a Rasch rating (or partial credit) model,
147 hindering the interpretation of the construct frequency(25, 26). One solution to this is to
148 collapse response categories, but it is preferable to present response options to participants
149 that avoid these issues altogether.

150

151 Despite these potential advantages, the use of dichotomous response options has been
152 questioned in other contexts, such as when measuring anxiety related to health and in
153 scales measuring attitudes and opinions (27, 28). There are concerns that a dichotomized
154 format of responses in scales discriminate less well between degrees of the construct of
155 interest (e.g., varying severity and frequency of experiencing anxiety). Furthermore, it is
156 unclear how respondents will answer when they have fewer response options, especially if
157 they only occasionally experience the condition under question. When offered only
158 dichotomous options, some might not consider an experience that occurred infrequently to
159 merit affirmation (i.e., they might respond “no/never” when the experience only occurred
160 rarely, e.g., once or twice). Therefore, despite dichotomous response options potentially
161 reducing complexity and respondent burden, they may compromise the scale's sensitivity to
162 varying frequencies of water insecurity experiences. This trade-off must be critically
163 evaluated in diverse settings.

164

165 Considering these potential advantages and disadvantages, it is uncertain how accurately
166 the WISE Scales with dichotomous responses would capture experiences of water
167 insecurity. Therefore, we sought to understand the potential consequences of using
168 dichotomous responses rather than polytomous response options for the WISE Scales.
169 Because the WISE scales have thus far primarily been administered using polytomous
170 response options, we addressed this goal by simulating dichotomous responses. We used
171 two dichotomization scenarios to account for any potential uncertainty around how people

172 who responded “rarely” to any of these experiences may respond when given a dichotomous
173 option. We used nationally representative datasets from 39 countries in which the 12-item
174 WISE Scales had been administered. Specifically, we sought to answer four questions:

- 175 1. Are “rarely” responses affirmed frequently enough to be meaningful in the
176 calculation of water insecurity experience scores?
- 177 2. Can 12-item WISE scores calculated using simulated dichotomous responses
178 accurately predict scores calculated using (the original) polytomous responses?
- 179 3. How does the predictive accuracy of the 12-item WISE Scales using simulated
180 dichotomous responses compare to that of the 4-item WISE Scales using
181 polytomous responses?
- 182 4. Is it possible to create four ordinal categories of water insecurity using simulated
183 dichotomous responses to the 12-item WISE scales that distinguish between
184 levels of water insecurity severity as well as the categories made using the
185 original scoring?

186

187 **Methods**

188 ***Study design, population, and key variables***

189 This analysis used nationally representative datasets from two sources: IWISE data from the
190 2020 and 2022 Gallup World Poll (GWP) (29, 30) and HWISE data from the 2021 Mexican
191 National Health and Nutrition Survey (ENSANUT 2021) (31). GWP implemented the IWISE
192 Scale with a 12-month recall period among individuals aged 15 years and older across 31
193 countries between September 2020 and February 2021 (30), and a further 7 countries in
194 2022 (n=50,768). The methodology for data collection and obtaining informed consent from
195 participants followed Gallup’s established protocols, which have been detailed elsewhere (6,
196 29) and approved by governing bodies as required in each country. ENSANUT 2021
197 surveyed 12,463 households, with a recall period of 4 weeks. Full details on the ENSANUT
198 survey methodology and nationally representative sampling strategy have been described
199 elsewhere (31). ENSANUT participants provided written informed consent.

200

201 GWP surveys included the WISE Scales and an item about water quality dissatisfaction. The
202 WISE Scales ask about 12 experiences related to problems with water, including modified or
203 limited behaviours (e.g., unable to wash hands), psychosocial impacts (e.g. worry about
204 water), and supply interruptions (Figure 1, full phrasing in Supplementary Table S1).
205 Respondents were asked to report how frequently they (when the IWISE Scale was used) or
206 anyone in their household (when the HWISE Scale was used) experienced these issues.
207 Data were collected with response options “never” (scored as 0), “rarely” (1), “sometimes”
208 (2), and “often” or “always” (3).

209

210 Dissatisfaction with water quality, a measure previously used to assess the construct validity
211 of the IWISE-12 scale (6), was measured with the item “In your city or area where you live,
212 are you satisfied or dissatisfied with the quality of water?”. Participants responded either
213 “satisfied” or “dissatisfied” that was coded as binary variable: (dissatisfied-1; satisfied-0).

214

215 ***Calculating WISE scores and categories using the 12-item scales with polytomous*** 216 ***response options***

217 WISE Scale responses are summed to create a score with a possible range of 0 to 36, with
218 higher scores indicating greater water insecurity. Cut-points have been established to
219 classify individuals and households as experiencing four levels of water insecurity: “no-to-
220 marginal” (scores of 0-2), “low” (3-11), “moderate” (12-23), or “high” water insecurity (24-36)
221 (6, 12).

222

223 ***Calculating WISE scores using the 12-item scales with dichotomized response*** 224 ***options: two scenarios***

225 Using the data from the polytomous scales described above, we simulated dichotomised
226 responses (i.e., participants affirming whether the experience occurred during the given
227 recall period or not). We simulated two potential scenarios. In the first scenario, termed “Any

228 Affirmation,” any affirmative response (“rarely,” “sometimes,” “often”, or “always”) was
229 recoded as “yes” (1), and “never” was recoded as “no” (0). Given the aforementioned
230 uncertainty about how individuals who responded “rarely” to an experience may answer
231 questions with dichotomous responses, we simulated a second scenario. In the second
232 scenario, termed “Sometimes-to-Always Affirmation”, “sometimes”, “often”, and “always”
233 were recoded as “yes” (1), whereas “rarely” and “never” were recoded as “no” (0). The
234 resulting summed scores for both versions ranged from 0 to 12.

235

236 ***Calculating WISE scores and categories using the 4-item scales with polytomous*** 237 ***responses***

238 The abbreviated 4-item versions of the IWISE and HWISE Scales, referred to as IWISE-4
239 (13) and HWISE-4 (14), respectively, include a subset of four of the 12 experiences asked
240 about in the full versions: worrying about not having enough water, not being able to wash
241 hands after dirty activities due to problems with water, not having enough water to drink, and
242 having to change plans due to problems with water (Figure 1). These responses are
243 summed to create overall water insecurity scores; these can range from 0-12. A score of ≥ 4
244 has been used as a cut-point to categorize individuals or households as experiencing water
245 insecurity (13, 14).

246

247 ***Statistical analysis using IWISE data***

248 For our first question (if “rarely” responses are affirmed sufficiently frequently to be
249 meaningful in the gradation of water insecurity), we plotted the response frequency for each
250 experience across all countries in the Gallup World Poll. To identify potential differences in
251 the frequency of affirming “rarely” by national water insecurity burden, we also plotted the
252 frequency among countries with low (i.e., United States, Australia) and high (i.e., Cameroon,
253 Zambia) national prevalence of water insecurity. We then used multiple logistic regression
254 models, adjusted for country, to test whether responding “rarely” to an experience was
255 associated with self-reported dissatisfaction with one’s water quality.

256

257 For our second question (the predictive accuracy of simulated dichotomous WISE responses
258 to the 12-item scales, relative to those with polytomous responses), we conducted four sets
259 of analyses. First, we used linear regression models to regress the scores from polytomous
260 response options on the scores from dichotomised response options. In these models, we
261 estimated root mean squared errors (RMSE) to quantify the magnitude of error due to
262 dichotomising responses. These models were estimated separately for each country, and
263 the average and ranges across countries were calculated

264

265 Second, we constructed receiver operator characteristic (ROC) curves to evaluate the
266 sensitivity and specificity of different cut-points for scores calculated using dichotomized
267 responses in relation to moderate-to-high water insecurity, as classified using scores
268 generated from the original scale with polytomous responses. We then examined the areas
269 under the curve (AUC) to understand the accuracy of the scores calculated using the two
270 dichotomous scenarios (“Any Affirmation” and “Sometimes-to-Always Affirmation”).

271

272 Third, we calculated the simulated weighted prevalence of moderate-to-high water insecurity
273 using the optimal cut-points identified by the ROC curves and compared these to the
274 estimated prevalence when using scores generated from polytomous responses. For each
275 country, we estimated the absolute percentage-point differences in prevalence estimates,
276 the percentage of people correctly classified, and the AUC.

277

278 Finally, we used logistic regression to compare how scores generated from dichotomized
279 and polytomous responses predicted water quality dissatisfaction. We compared a 3-point
280 difference in the scores from polytomous to a 1-point difference in the dichotomized
281 responses versions. We examined the AUC to compare the accuracy of the polytomous
282 compared to the dichotomized versions. We also ran logistic regressions using the optimal
283 cut-points identified by the ROC curves for dichotomized versions, and a cut-point of ≥ 12 for

284 the polytomous version to examine how these definitions of moderate-to-high water
285 insecurity were associated with the odds of water quality dissatisfaction.

286

287 For our third question (about the predictive accuracy of the WISE Scales using dichotomized
288 responses compared to that of the 4-item WISE Scales using polytomous responses), we
289 ran linear regressions to test the association between scores from the 12-item scales (using
290 polytomous and dichotomous responses) and scores from the 4-item scale with polytomous
291 responses. We calculated RMSE and residuals to quantify prediction errors. Additionally, we
292 calculated the weighted prevalence of moderate-to-high water insecurity using a cut-point of
293 ≥ 4 for the 4-item scale (13) and compared it to that estimated using the original 12-item
294 scale. We then used logistic regression to understand if water insecurity, as classified using
295 the four different versions of the tool, was associated with water quality dissatisfaction, which
296 was used to assess the construct validity of the IWISE-12 scale (6).

297

298 For our fourth question (if it is possible to calculate 4 levels of water insecurity using scores
299 generated from dichotomized response options), we used ROC curves to determine whether
300 cut-points in scores from the dichotomous responses could be identified. Specifically, we
301 tried to determine if we could identify cut-points with high sensitivity and specificity for each
302 water insecurity category, as calculated using scores from the full scale with polytomous
303 responses.

304

305 ***Sensitivity analysis using HWISE data***

306 To assess whether the answers to our four research questions differed when using
307 household-level observations, we repeated these analyses with data from ENSANUT 2021,
308 which is, to our knowledge, the only survey vehicle through which nationally representative
309 HWISE data have been collected. A total of 12,619 households were interviewed using the
310 HWISE module, of which 156 households were missing responses to one or more WISE

311 experiences. These households were excluded from the analysis, resulting in a final sample
312 of 12,463 households.

313
314 Associations between WISE scores calculated using dichotomized compared to polytomous
315 response were estimated using linear regression. We used ROC curves to explore cut-points
316 in the scores generated from dichotomized responses that maximized sensitivity and
317 specificity for classifying moderate-to-high water insecurity, as determined using WISE
318 scores from polytomous responses (3). Prevalence estimates (unweighted) of moderate-to-
319 high household water insecurity, as assessed using the identified optimal cut-points for the
320 scores calculated with dichotomous responses, were compared to that estimated using
321 scores calculated with polytomous responses. A question about water quality satisfaction
322 was not asked in this survey.

323

324 ***Ethics***

325 This study using secondary, deidentified data was determined to not constitute human
326 subjects research by the Institutional Review Board at the University of North Carolina at
327 Chapel Hill. Gallup World Poll survey procedures were approved by governing bodies as
328 required in each country. Gallup obtained informed consent from participants. We received
329 deidentified data from Gallup for our analyses. All ENSANUT survey procedures were
330 reviewed and approved by the Research, Biosecurity, and Ethics Committees of the National
331 Institute of Public Health, Mexico. Each respondent to the household survey provided his or
332 her written informed consent (Project ID: 1750).

333

334 **Results**

335 The IWISE Scale was administered to 52,560 individuals in 38 countries through the Gallup
336 World Poll. Of these, 1,792 (3.4%) were missing values to one or more experiences and
337 excluded from the analysis using the full scores, resulting in a final sample of 50,768
338 individuals.

339

340 ***The contribution of “rarely” to WISE scores (Question 1)***

341 “Rarely” was a common response to each experience (**Figure 2A**). For each item, between
342 32% and 37% of respondents who affirmed an experience reported it as occurring “rarely”
343 (**Supplementary Table S2**). In countries with a low prevalence of moderate-to-severe water
344 insecurity, such as Australia (A, 0.973%) and the US (U, 3.67%), “rarely” accounts for most
345 of the affirmations (**Figure 2B**). In Australia, between 52% to 93% of respondents who
346 affirmed an experience reported it as occurring “rarely”, compared to between 58% and 79%
347 in the United States (Supplementary Table S2). Conversely, for countries experiencing a
348 high national prevalence of moderate-to-high water insecurity, such as Cameroon (C,
349 63.9%) and Zambia (Z, 48.1%), “rarely” was affirmed less frequently (**Figure 2C**). In
350 Cameroon, between 19% to 27% of respondents affirming each experience reported it as
351 occurring “rarely”, compared to 23% to 29% in Zambia (Supplementary Table S2).

352

353 In logistic regression for each WISE item, the odds of reporting water quality dissatisfaction
354 increased monotonically across polytomous options (**Table 1**). The odds of reporting water
355 quality dissatisfaction were higher among those who responded “rarely” compared to those
356 who responded “never” experiencing a given issue. These results suggest that rare
357 occurrences of water issues can meaningfully predict other constructs related to water
358 insecurity.

359

360 ***Figure 2. Distribution of responses to water insecurity experiences (unweighted) in***
361 ***nationally representative data from 38 countries (n= 50,768, Gallup World Poll 2020, 2023).***
362 ***A: Aggregated across countries. B: Distribution of “rarely” responses in countries with a***
363 ***relatively low prevalence of moderate-to-high water insecurity (Australia (A) 0.973%; USA***
364 ***(U) 3.67%). C: Distribution of “rarely” responses in countries with a relatively high prevalence***
365 ***of moderate-to-high water insecurity (Cameroon (C) 63.9%; Zambia (Z) 48.1%).***

366

367 **Table 1.** Odds of reporting water quality dissatisfaction in relation to the reported frequency
368 of experiencing each WISE item (weighted and adjusted for country) using nationally
369 representative data from 38 countries (n= 50,768, Gallup World Poll 2020, 2022).*

370 *Reference for all models was “Never”.

371

372 **The predictive accuracy of scores calculated using dichotomized responses**

373 **(Questions 2 & 3)**

374 Both versions of the dichotomised-response scores – “Any Affirmation” and “Sometimes-to-
375 Always Affirmation” – accurately predicted scores calculated using the 12-item IWISE Scale
376 with polytomous responses (**Table 2**; see Supplementary Tables S3A and S3B for country-
377 specific results).

378

379 The RMSE (i.e., the standard deviation of the residuals from the regression model) was
380 lower for the “Sometimes-to-Always Affirmation” version compared to the “Any Affirmation”
381 version, indicating better overall predictive accuracy of the former. There was, however,
382 greater variability of the residuals at higher values of the “Sometimes-to-Always Affirmation”
383 version, whereas the variability of the residuals appeared to be even across values of the
384 “Any Affirmation” (Supplementary Figure S1); both exhibited heteroskedasticity. The residual
385 pattern and RMSE of the 4-item IWISE Scale were similar to that of the “ Sometimes-to-
386 Always Affirmation” version (**Table 2**; see Supplementary Table S3C for country-level
387 results). The mean beta coefficient from the models regressing the polytomous score on the
388 dichotomized scores was highest for the IWISE-4 Scale (2.65, range: 2.24-2.84) and lowest
389 for the “Any Affirmation” version (1.95, range: 1.39-2.26). The “ Sometimes-to-Always
390 Affirmation” version showed a slightly higher mean beta value of 2.35, ranging from 2.13 to
391 2.60 across countries. The IWISE-4 scale had the highest mean beta value of 2.68, ranging
392 from 2.24 to 2.84 across countries. Despite these differences, the R-squared values,
393 correlation coefficients, and standard errors were similar across both dichotomized versions

394 and IWISE-4. In short, both dichotomized versions and IWISE-4 had similar predictive
395 accuracy.

396

397 **Table 2.** Unweighted linear regression of the 12-item IWISE Scale (using polytomous
398 responses) on simulated IWISE scores (using two strategies for dichotomizing responses,
399 “Any Affirmation” and “Sometimes-to-Always Affirmation”) and scores from the 4-item IWISE
400 Scale, averaged across 38 countries (n=50,768*, Gallup World Poll 2020, 2022)

401 **Per country N Mean=980; Median=1336; Range= 878-12349; **RMSE: Root Mean
402 Squared Error.*

403

404 The AUC for scores calculated using both versions of dichotomization showed high
405 accuracy, with values close to 0.98 (Supplementary Figure S2). This indicates that both
406 versions were accurate at predicting moderate-to-high water insecurity, as defined as scores
407 ≥ 12 in the original 12-item scale with polytomous responses.

408

409 For the "Sometimes-to-Always Affirmation" dichotomised version, cut-points of ≥ 4 and ≥ 5
410 resulted in the highest overall correct classification of moderate-to-high water insecurity, at
411 94% and 95%, respectively (Supplementary Tables S4A and S4B). For the "Any Affirmation"
412 dichotomised version, greater accuracy was achieved with higher cut-points (≥ 6 and ≥ 7),
413 although a high percentage were correctly classified when using a cut-point of ≥ 4 (86%) and
414 ≥ 5 (90%). Country-specific details on the percentage correctly classified and AUC for both
415 versions are available in Supplementary Tables S5A and S5B.

416

417 We identified different cut-points for estimating water insecurity prevalence using the two
418 dichotomized versions (**Figure 3**). A cut-point of ≥ 4 in the “Sometimes-to-Always Affirmation”
419 version provided an estimate of water insecurity prevalence that was 2 percentage points
420 higher than the estimate from polytomous responses, whereas that same cut-point in the
421 “Any-Affirmation” version resulted, on average, in a 16 percentage-point over-estimation

422 **(Figure 3)**. In contrast, a cut-point of ≥ 7 for “Any-Affirmation” resulted in a 1 percentage-
423 point over-estimation of water insecurity prevalence but a 12 percentage-point under-
424 estimation of water insecurity prevalence using the “Sometimes-to-Always Affirmation”
425 version. (Tables S6A and S6B in the supplementary files show the weighted prevalence
426 estimates per cut-point for each country.) By comparison, using a cut-point of ≥ 4 for IWISE-4
427 (which has been previously validated), resulted in an average 3-percentage-point
428 overestimation (Supplementary Table S6C). Therefore, while similar prevalences can be
429 estimated using both dichotomized versions, the cut-points will differ depending on whether
430 response patterns with dichotomous options align more closely with the simulated “Any
431 Affirmation” version or “Sometimes-to-Always Affirmation” version.

432

433 **Figure 3.** Average absolute differences in prevalence estimates of moderate-to-high water
434 insecurity, comparing the estimated prevalence from the 12-item scale with polytomous
435 responses to those estimated using various cut-points with the “Any Affirmation” and
436 “Sometimes-to-Always Affirmation” versions (weighted), based on nationally representative
437 data from 38 countries ($n = 50,768$, Gallup World Poll 2020, 2023).

438

439 The associations between dichotomised-response IWISE scores and odds of reporting water
440 quality dissatisfaction (**Figure 4**) were similar for both the “Any Affirmation” (red line) and “
441 Sometimes-to-Always affirmation” (green line) scenarios compared to that observed when
442 using polytomous responses (blue line). The 4-item IWISE Scale (yellow line) had
443 comparable associations, demonstrating the consistency of these results across different
444 scoring methods. Similar AUC values suggests that both dichotomized versions and IWISE-
445 4 had comparable accuracy to the original 12-item scale with polytomous responses in
446 predicting dissatisfaction with water quality. In other words, the abbreviated scales behaved
447 similarly to the full scale in predicting another construct related to water insecurity.

448

449 **Figure 4.** Predicted probability of reporting dissatisfaction with water quality by each IWISE
450 response score option. Score options include the full IWISE Scale using polytomous
451 responses (at 3-point intervals), the full IWISE Scale using dichotomized responses (“Any
452 Affirmation” and “Sometimes-to-Always Affirmation” version), and the 4-item IWISE Scale
453 using polytomous responses, based on nationally representative data from 38 countries (n=
454 50,768, Gallup World Poll 2020, 2022).*

455 *All models were adjusted for country and weighted by survey weights. For the full IWISE12
456 score with polytomous items, the range was 0 to 36 (i.e., each point as labeled on the x-axis
457 corresponds to 3 points on the full IWISE12 polytomous item score)

458

459 When using various cut-points (≥ 4 , ≥ 5 , ≥ 6 , and ≥ 7) to define moderate-to-high water
460 insecurity, both dichotomized versions yielded odds of reporting water quality dissatisfaction
461 that were similar to those obtained when using a cut-point of ≥ 12 for the full scale with
462 polytomous responses (Supplementary Table S7). For example, for individuals experiencing
463 moderate-to-high water insecurity, as classified using the full scale with polytomous
464 responses, the odds of water quality dissatisfaction were 4.5 times higher (95% CI: 4.18-
465 4.85) compared to those experiencing no-to-low water insecurity. In comparison, the
466 estimated odds for the "Any-Affirmation" version at a cut-point of ≥ 5 was 4.28 (95% CI: 3.99-
467 4.58) and 4.11 (95% CI: 3.81-4.44) for the same cut-point using the "Sometimes-to-Always
468 affirmation" version. At higher cut-points, similar associations were observed, although the
469 strength of associations tended to decrease as the cut-points increased. The IWISE-4 scale,
470 using a cut-point of ≥ 4 , had a comparable association with water quality dissatisfaction (OR:
471 4.29, 95% CI: 3.99-4.61). These results indicate that all scoring versions exhibited similar
472 construct validity.

473

474 **Creating water insecurity categories with dichotomized responses (Question 4)**

475 Using ROC curves, we identified cut-points that enabled the categorization of water
476 insecurity for both dichotomized versions ("Any" " and "Sometimes-to-Always" affirmations).

477 These cut-points resulted in similar distributions of individuals across water insecurity
478 categories (no-to-marginal, low, moderate, and high) when compared to the 12-item scale
479 with polytomous responses (**Figure 5**). Thus, it was possible to estimate ordinal water
480 insecurity categories using both dichotomized versions. (Supplementary Tables S8A-S9B
481 provide details on the proportion of individuals correctly classified at each cut-point, as well
482 as the AUC values for countries with low and high overall water insecurity.)

483 These results demonstrate the feasibility of categorizing water insecurity using scores
484 calculated with dichotomous response options. We cannot propose definitive cut-offs,
485 however, because the current analysis relies on simulated data. Establishing appropriate
486 cut-offs requires empirical data that capture how individuals respond when the items are
487 explicitly presented with dichotomous response options.

488

489 ***Figure 5.** Proportion of individuals classified within each level of water insecurity based on*
490 *the validated cut-offs for the original 12-item IWISE Scale with polytomous responses, and*
491 *the cut-offs identified for the dichotomized versions (“Any-Affirmation” and “Sometimes-to-*
492 *Always affirmation”)* based on nationally representative data from 38 countries ($n= 50,768$,
493 *Gallup World Poll 2020, 2022).*

494

495 **Sensitivity analyses using HWISE data**

496 We observed similar relationships using HWISE data collected in the ENSANUT survey in
497 Mexico (Supplementary Text 1, Supplementary Tables S10-S13, Supplementary Figures S3-
498 S5).

499

500 **Discussion**

501 We evaluated whether experiencing issues with water access and use only rarely was
502 associated with other water problems (e.g., dissatisfaction with water quality) and simulated
503 the potential consequences of administering WISE scales with dichotomous instead of
504 polytomous responses using data from nationally representative surveys in 39 countries (see

505 **Table 3).** First, even a rare experience of any of the 12 WISE items was strongly associated
506 with higher odds of also reporting dissatisfaction with water quality. Second, two scenarios
507 for dichotomizing polytomous items (either considering rare experiences as an affirmation or
508 not) were simulated. Both dichotomization scenarios accurately predicted the 12-item scale
509 score calculated using polytomous responses. Third, the predictive accuracies of the 12-item
510 scale with dichotomized responses were similar to the predictive accuracy of the abbreviated
511 IWISE-4. Fourth, estimation of low, moderate, and high water-insecurity categories was
512 reasonable with both versions of the dichotomized WISE scales. Taken together, these
513 findings suggest that administering the WISE scales with dichotomous instead of polytomous
514 response options may be a useful strategy in some situations, but it may come at the cost of
515 some lost information, as discussed below.

516

517 **Table 3. Summary of research questions, analyses performed, and results**

518

519 Given the importance of even “rare” experiences in the gradation of water insecurity
520 experiences, the interpretation of water insecurity may be affected if rare occurrences are
521 not adequately captured when items are administered with dichotomous response options
522 (i.e., if respondents chose not to affirm an experience that occurred only once or twice).
523 Thus, when administering the WISE Scales with dichotomous response options, efforts
524 should be made to encourage respondents to carefully consider if these experiences have
525 occurred even once over the recall period, and if so, to affirm these experiences, so as not to
526 miss people who are experiencing infrequent water insecurity. That is, instructions must
527 clarify that any occurrence should be considered an affirmation.

528

529 Dichotomous response options may limit the ability to understand water insecurity dynamics
530 in some situations. For example, an intervention may cause people to shift from
531 experiencing an issue often to rarely, as has been found for food insecurity (32). This

532 difference is meaningful for understanding the impact of an intervention but would be lost if
533 only dichotomous response options were provided. We therefore recommend polytomous
534 response options that capture the frequency of experience for program evaluations.
535 Similarly, understanding the frequency with which these experiences occur might be
536 important for designing effective targeted strategies, as those experiences issues more
537 frequently may require a different level of intervention, understanding frequency of food
538 insecurity coping strategies has been shown to be useful to inform targeting of food security
539 interventions (33).

540

541 Both versions of simulated dichotomized responses (i.e., whether rare experiences were
542 considered an affirmation or not) accurately predicted IWISE and HWISE scores calculated
543 using the polytomous response options. The scores from dichotomized responses also had
544 similar accuracy in predicting a related construct of water insecurity (dissatisfaction with
545 water quality). The WISE-4 has similar predictive accuracy to the dichotomized scores, both
546 in terms of predicting scores calculated from WISE-12 polytomous responses and predicting
547 a related water insecurity construct. While both abbreviated versions of the scales (WISE-4
548 and WISE-12 with dichotomous responses) may offer viable alternatives to the full scale
549 when time and resources are limited, the 12-item WISE scales with dichotomous responses
550 will better capture the full array of ways in which water insecurity can manifest and interrupt
551 life.

552

553 The prevalence of no-to-marginal, low, moderate, and high water insecurity could be
554 estimated using WISE scores with dichotomized responses, providing a clear advantage
555 over the abbreviated IWISE-4 and HWISE-4 Scales. What those cut-points are, however, will
556 depend on how participants respond when offered dichotomous response options. For
557 IWISE, if all respondents who responded “rarely” were considered to have affirmed the
558 experience (the “Any Affirmation” scenario), then a cut-point of ≥ 7 has the best specificity
559 and a cut-point of ≥ 6 has the best sensitivity for classifying moderate-to high water

560 insecurity. If, however, those who responded “rarely” were not to have not affirmed the
561 experience (the “Sometimes-to-Always Affirmation” scenario), then a cut-point of ≥ 5 has the
562 best specificity and a cut-point of ≥ 4 has the best sensitivity for classifying moderate-to high
563 water insecurity. Ultimately, the establishment of appropriate thresholds for defining different
564 levels of water insecurity with WISE scales scores with dichotomized responses must be
565 based on data from WISE surveys administered with this response format from the start.

566

567 A challenge in knowing which of the versions of simulated dichotomized responses best
568 illustrates the amount of information lost when administering WISE surveys with
569 dichotomized responses is that we do not know how people would have responded if they
570 had been presented with a dichotomous response option. Research is needed in which
571 dichotomous responses are presented to the respondent, ideally in direct comparison using
572 a split sample to the presentation of polytomous responses. One such split-sample study
573 has been conducted in Mexico with HWISE-12; this study found that the prevalences of
574 water insecurity estimated in the sub-sample that was administered the survey with
575 dichotomous response options was comparable to the prevalence of the sub-sample that
576 was administered the survey with polytomous response options (21).

577

578 Further research is required to understand how affirmations might change depending on
579 many response options are presented and whether prompts are provided to encourage
580 respondents to consider rare occurrences as an affirmation. Further research is also
581 required to understand how respondents’ affirmation of items that occur only rarely might
582 change when presented with dichotomous (experienced the issue or not) instead of a
583 polytomous (frequency of experiencing the issue) response options. It will also be important
584 to assess if losing this nuance is worth the practical, logistical, or cost advantages.
585 Specifically, comparing findings from administering the scales with dichotomous and
586 polytomous response options in similar populations will permit informed decisions about
587 which format of responses best capture the information that is most important to

588 organizations, researchers, and policymakers. It will also be valuable to validate
589 dichotomous response options in diverse contexts, particularly high-income countries and
590 areas with low water insecurity prevalence, to determine their robustness across settings.
591 Whilst the sample used in our study contained data from low-, middle-, and high- income
592 countries, there are only two countries that were formally classified as high-income. Further
593 research is required in high-income settings to understand how use of dichotomous
594 response options in WISE surveys may affect measurement of water insecurity in contexts
595 where it is less prevalent.

596

597 **Conclusion**

598 Polytomous responses options provide more information, but dichotomous response options
599 hold promise for measuring water insecurity when there is need for a more rapid but still
600 comprehensive survey. For program evaluation, WISE Scales with polytomous responses
601 are likely more suitable because they offer greater nuance in understanding both if an
602 experience occurred as well as its frequency. WISE Scale items with dichotomous response
603 options have the potential to provide a time-saving, valid alternative to polytomous response
604 options for measuring occurrence of water insecurity experiences and estimating prevalence
605 of no-to-marginal, low, moderate, and high water insecurity. Additional data collection using
606 dichotomously phrased responses is needed to fully understand all that might be gained and
607 lost with the dichotomization of WISE response options.

608

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621

622 **Data Availability**

623 This study was based on survey data collected by Gallup World Poll (GWP) and ENSANUT
624 2021. Data from ENSANUT are publicly available on the website of the National Institute of
625 Public Health (Mexico). Nationally aggregated WISE data can be accessed
626 from <https://doi.org/10.21985/n2-avk4-9932>, additional data can be requested by
627 emailing WISE_scales@northwestern.edu.

628 **References**

629 1. Jepson WE, Wutich A, Collins SM, Boateng GO, Young SL. Progress in household water
630 insecurity metrics: a cross-disciplinary approach. *WIREs Water*. 2017;4(3):e1214.

631

632 2. Rosinger AY, Young SL. The toll of household water insecurity on health and human
633 biology: Current understandings and future directions. *WIREs Water*. 2020;7(6):e1468.

634

635 3. Young SL, Boateng GO, Jamaluddine Z, Miller JD, Frongillo EA, Neilands TB, et al. The
636 Household Water InSecurity Experiences (HWISE) Scale: development and validation of a
637 household water insecurity measure for low-income and middle-income countries. *BMJ*
638 *Global Health*. 2019;4(5):e001750.

639

640 4. Young SL, Frongillo EA, Jamaluddine Z, Melgar-Quiñonez H, Pérez-Escamilla R, Ringler
641 C, Rosinger AY. Perspective: The Importance of Water Security for Ensuring Food Security,
642 Good Nutrition, and Well-being. *Advances in Nutrition*. 2021;12(4):1058-73.

643

644 5. Young SL, Miller, Joshua D, Bose, Indira. Measuring human experiences to advance safe
645 water for all. Evanston, Illinois: Institute for Policy Research, Northwestern University; 2024.

646

647 6. Young SL, Bethancourt HJ, Ritter ZR, Frongillo EA. The Individual Water Insecurity
648 Experiences (IWISE) Scale: reliability, equivalence and validity of an individual-level
649 measure of water security. *BMJ Global Health*. 2021;6(10):e006460.

650

651 7. Slaymaker T, Johnston, R., Young, S.L., Miller, J., & Staddon, C. Detailed Review of a
652 Recent Publication: An innovative measure of the experience of water insecurity can help
653 identify vulnerable households and evaluate interventions. *UNC Water Institute- WaSH*
654 *Policy Research Digest*; 2020.

655

- 656 8. Young SL, Bose, I., Miller, J. D., Pascavis, K., Alzarez, K., Barstow, C., Boller M, Brogan,
657 J., Collins, S., Connors, K., Durham, T., Frongillo, E., A. G-R, P., García, O. P., Marks, S.,
658 Melgar-Quiñonez, H., Miller, S. M-R, V., Muñoz-Espinosa, A., Ndege, G., Otieno, I, Pérez-,
659 Escamilla R, Rodas, S., Salles-Costa, R., Sarker, M. R., Shamah-Levy, T., Thuo S,
660 Webster, J., Yeye, V The Water Insecurity Experiences (WISE) Scales: A Manual for
661 Implementation and Analysis of People's Experiences with Water. Evanston, Illinois: Institute
662 for Policy Research, Northwestern University; 2024.
- 663
- 664 9. Editorial N. Water Crisis: How Local Technologies Can Help Solve a Global Problem.
665 Nature 620 (7972): 7. 2023.
- 666
- 667 10. Mundo-Rosas V, Shamah-Levy T, Muñoz-Espinosa A, Hernández-Palafox C, Vizuet-
668 Vega NI, de los Ángeles Torres-Valencia M, et al. Inseguridad alimentaria y del agua. salud
669 pública de méxico. 2024;66(4, jul-ago):581-8.
- 670
- 671 11. Melgar-Quiñonez H, Gaitán-Rossi P, Pérez-Escamilla R, Shamah-Levy T, Teruel-
672 Belismelis G, Young SL. A declaration on the value of experiential measures of food and
673 water insecurity to improve science and policies in Latin America and the Caribbean.
674 International Journal for Equity in Health. 2023;22(1):184.
- 675
- 676 12. Frongillo EA, Bethancourt HJ, Miller JD, Young SL. Identifying ordinal categories for the
677 Water Insecurity Experiences Scales. Journal of Water, Sanitation and Hygiene for
678 Development. 2024:washdev2024042.
- 679
- 680 13. Bethancourt HJ, Frongillo EA, Young SL. Validity of an abbreviated Individual Water
681 Insecurity Experiences (IWISE-4) Scale for measuring the prevalence of water insecurity in
682 low- and middle-income countries. Journal of Water, Sanitation and Hygiene for
683 Development. 2022;12(9):647-58.

684

685 14. Young SL, Miller JD, Frongillo EA, Boateng GO, Jamaluddine Z, Neilands TB. Validity of
686 a Four-Item Household Water Insecurity Experiences Scale for Assessing Water Issues
687 Related to Health and Well-Being. *Am J Trop Med Hyg.* 2021;104(1):391-4.

688

689 15. Bethany A. Caruso JC, Julie Hennegan, , Albert Motivans LP, Madeleine Patrick, Beesan
690 , Shonnar SS, Nicole Stephan. Priority Gender-Specific Indicators for WASH Monitoring
691 under SDG Targets 6.1 and 6.2: Recommendations for National and Global Monitoring. New
692 York: United Nations Children’s Fund (UNICEF) and World Health Organization (WHO);
693 2024.

694

695 16. Young S, Bethancourt H, Frongillo E, Bose I. Water Insecurity Experiences (WISE)
696 Country Profiles. Evanston, Illinois. : Institute for Policy Research, Northwestern University;
697 2024.

698

699 17. Pérez-Escamilla R, Salles-Costa R, Segall-Corrêa AM. Food insecurity experience-
700 based scales and food security governance: A case study from Brazil. *Global Food Security.*
701 2024;41:100766.

702

703 18. Pérez-Escamilla R. Food and nutrition security definitions, constructs, frameworks,
704 measurements, and applications: global lessons. *Frontiers in Public Health.*
705 2024;12:1340149.

706

707 19. Salles-Costa R, Segall-Corrêa AM, Alexandre-Weiss VP, Pasquim EM, Paula NMd,
708 Lignani JdB, et al. Rise and fall of household food security in Brazil, 2004 to 2022. *Cadernos*
709 *de Saúde Pública.* 2023;39:e00191122.

710

- 711 20. Pérez-Escamilla R, Vilar-Compte M, Gaitan-Rossi P. Why identifying households by
712 degree of food insecurity matters for policymaking. *Global Food Security*. 2020;26:100459.
713
- 714 21. García OP CM, Parás P, Melgar-Quiñonez H. . Validity of a binary Household Water
715 Insecurity Scale for assessing water-related issues and its relationship with public health
716 indicators. Under review. 2024.
717
- 718 22. Engelhard Jr G, Wang J. Rasch models for solving measurement problems: Invariant
719 measurement in the social sciences: Sage Publications; 2021.
720
- 721 23. Cafiero C, Nord M, Viviani S, Del Grossi ME, Ballard T, Kepple A, et al. Methods for
722 estimating comparable prevalence rates of food insecurity experienced by adults throughout
723 the world. Rome: Food and Agriculture Organization of the United Nations (FAO); 2016
724 2016.
725
- 726 24. Cafiero C, Viviani S, Nord M. Food security measurement in a global context: The food
727 insecurity experience scale. *Measurement*. 2018;116:146-52.
728
- 729 25. De Ayala RJ. The theory and practice of item response theory: Guilford Publications;
730 2013.
731
- 732 26. Bond TG, Fox CM. Applying the Rasch model: Fundamental measurement in the human
733 sciences: Psychology Press; 2013.
734
- 735 27. Welch PG, Carleton RN, Asmundson GJG. Measuring health anxiety: Moving past the
736 dichotomous response option of the original Whiteley Index. *Journal of Anxiety Disorders*.
737 2009;23(7):1002-7.
738

- 739 28. Preston CC, Colman AM. Optimal number of response categories in rating scales:
740 reliability, validity, discriminating power, and respondent preferences. *Acta Psychologica*.
741 2000;104(1):1-15.
742
- 743 29. Gallup. *Worldwide research methodology and codebook*. Washington, DC; 2020.
744
- 745 30. Young SL, Bethancourt HJ, Ritter ZR, Frongillo EA. Estimating national, demographic,
746 and socioeconomic disparities in water insecurity experiences in low-income and middle-
747 income countries in 2020–21: a cross-sectional, observational study using nationally
748 representative survey data. *The Lancet Planetary Health*. 2022;6(11):e880-e91.
749
- 750 31. Romero-Martínez M, Barrientos-Gutiérrez T, Cuevas-Nasu L, Bautista-Arredondo S,
751 Colchero MA, Gaona-Pineda EB, et al. National Health and Nutrition Survey 2021
752 methodology. *Salud Pública de México*. 2021;63(6):813-8.
753
- 754 32. Cohen CR, Weke E, Frongillo EA, Sheira LA, Burger R, Mocello AR, et al. Effect of a
755 multisectoral agricultural intervention on HIV health outcomes among adults in Kenya: a
756 cluster randomized clinical trial. *JAMA Network Open*. 2022;5(12):e2246158-e.
757
- 758 33. Maxwell D, Watkins B, Wheeler R, Collins G. *The coping strategies index: A tool for*
759 *rapidly measuring food security and the impact of food aid programs in emergencies*.
760 Nairobi: CARE Eastern and Central Africa Regional Management Unit and the World Food
761 Programme Vulnerability Assessment and Mapping Unit. 2003.

763 **Table 1.** Odds of reporting water quality dissatisfaction in relation to the reported frequency
 764 of experiencing each WISE item (weighted and adjusted for country) using nationally
 765 representative data from 38 countries (n= 50,768, Gallup World Poll 2020, 2022).*

Items		Odds Ratio (OR)	95% Confidence Interval (CI)	
			Lower CI	Upper CI
Worry (n=51,941)	Rarely	2.92	2.67	3.19
	Sometimes	3.55	3.26	3.86
	Often/Always	8.20	7.40	9.08
Plans (n=51,875)	Rarely	2.84	2.59	3.11
	Sometimes	3.30	3.03	3.58
	Often/Always	6.08	5.39	6.86
Hands (n=52,014)	Rarely	2.27	2.02	2.55
	Sometimes	2.74	2.47	3.04
	Often/Always	4.29	3.65	5.04
Drink (n=52,009)	Rarely	2.62	2.36	2.90
	Sometimes	3.25	2.96	3.58
	Often/Always	5.07	4.39	5.86
Food (n=51,861)	Rarely	2.43	2.19	2.70
	Sometimes	3.12	2.86	3.41
	Often/Always	5.42	4.72	6.21
Interrupt (n=51,752)	Rarely	2.35	2.16	2.56
	Sometimes	3.11	2.86	3.38
	Often/Always	5.93	5.37	6.55
Body (n=52,028)	Rarely	2.38	2.15	2.64
	Sometimes	3.04	2.76	3.36
	Often/Always	4.99	4.28	5.83
Clothes (n=51,969)	Rarely	2.56	2.33	2.80
	Sometimes	3.17	2.90	3.45
	Often/Always	5.99	5.30	6.77
Angry (n=51,940)	Rarely	2.78	2.54	3.05
	Sometimes	3.45	3.18	3.75
	Often/Always	7.12	6.38	7.94
Sleep (n=52,013)	Rarely	2.69	2.39	3.03
	Sometimes	2.88	2.58	3.21
	Often/Always	4.85	4.05	5.79
None (n=51,969)	Rarely	2.71	2.45	3.00
	Sometimes	3.43	3.12	3.76
	Often/Always	6.04	5.25	6.94
Shame (n=51,897)	Rarely	2.55	2.30	2.82
	Sometimes	3.31	3.03	3.62
	Often/Always	5.62	4.91	6.43

766
 767 *Reference for all models was “Never”.

768 **Table 2.** Unweighted linear regression of the 12-item IWISE Scale (using polytomous
 769 responses) on simulated IWISE scores (using two strategies for dichotomizing responses,
 770 “Any Affirmation” and “Sometimes-to-Always Affirmation”) and scores from the 4-item IWISE
 771 Scale, averaged across 38 countries (n=50,768*, Gallup World Poll 2020, 2022)
 772

	Any Affirmation			Sometimes-to-Always Affirmation			4-item IWISE Scale		
	Mean	Median	Range	Mean	Median	Range	Mean	Median	Range
RMSE**	2.96	2.82	0.78-4.21	2.73	2.51	1.13-3.35	2.72	2.58	0.96-3.54
Beta coefficient	1.98	1.95	1.39-2.26	2.35	2.35	2.13-2.6	2.68	2.65	2.24-2.84
SE	0.0240	0.0236	0.00643-0.0353	0.0260	0.0254	0.00761-0.0463	0.0298	0.0302	0.00843-0.0371
R-Squared	0.864	0.862	0.747-0.932	0.884	0.894	0.725-0.939	0.882	0.883	0.804-0.954
Correlation	0.929	0.929	0.865-0.966	0.940	0.946	0.852-0.969	0.939	0.940	0.897-0.977

773
 774 *Per country N Mean=980; Median=1336; Range= 878-12349; **RMSE: Root Mean
 775 Squared Error.

776 **Table 3. Summary of research questions, analyses performed, and results**

Abbreviated research question	Analyses	Results
1. Is the response “rarely” meaningful in the gradation water insecurity experiences?	Estimated the frequency of people responding “rarely” to each of the WISE items. Tested if responses of “rarely” on different items predicted dissatisfaction with water quality using logistic regression.	Rarely experiencing a water related issue is strongly related to higher odds of dissatisfaction with water quality. (Figure 2, Table 1).
2. Do WISE-12 scores calculated with dichotomized responses accurately predict WISE scores calculated from polytomous responses?	2.1 Ran linear regression models, with dichotomous response scores as explanatory variable & polytomous response scores as outcome variable 2.2 Receiver operating characteristic (ROC) curves to explore the how well different scores using dichotomised responses accurately estimate moderate-to-high water insecurity 2.3 Estimated and compared prevalence of water insecurity using WISE scores with polytomous and dichotomous response options. 2.4 Compared how WISE scores from polytomous versus dichotomized responses predicted dissatisfaction with water quality using logistic regression models	WISE-12 scores from dichotomized responses provided a reasonable approximation to scores with polytomous responses and were similarly predictive of water quality dissatisfaction (Table 2, Supplementary Figure S2, Figure 4, Figure 5)
3. Do WISE-4 scores calculated with polytomous responses more accurately predict WISE-12 scores calculated from polytomous responses vs. WISE-12 scores calculated with dichotomous responses?	We repeated 2.1, 2.3, 2.4, using polytomous responses to 4 WISE items	WISE-12 scores from dichotomized responses and WISE-4 scores from polytomous responses provide comparable approximation to the WISE-12 scores with polytomous responses and are comparatively predictive of water quality dissatisfaction (Table 2, Supplementary Table S6C, Figure 5).
4. How well can different cut-offs of WISE scores from dichotomized responses differentiate between different levels of water insecurity estimated using WISE-12 scores with polytomous responses?	Used ROC to find cutoffs that approximate categories of no-to-marginal, low, moderate, or high water insecurity as defined from WISE-12 scores using polytomous responses.	WISE scores from the dichotomized responses can be used to estimate the number of individuals experiencing no-to-marginal, low, moderate, or high water insecurity. Exact cut-offs should be developed and validated using data collected with dichotomous responses rather than using dichotomized data from polytomous responses.

777



Figure 1

A

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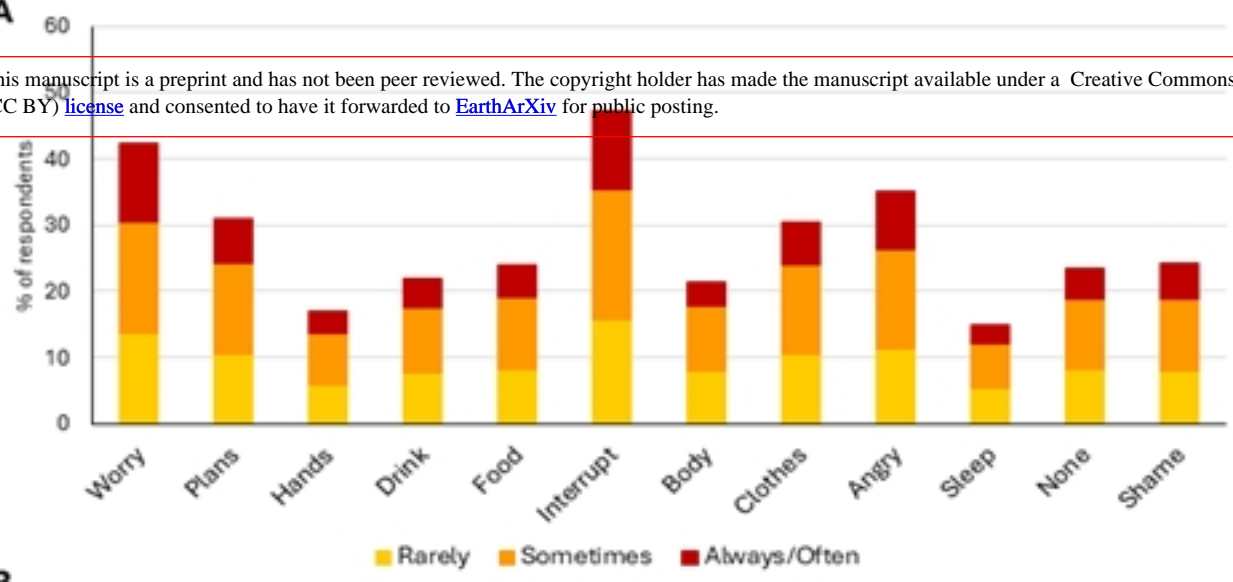
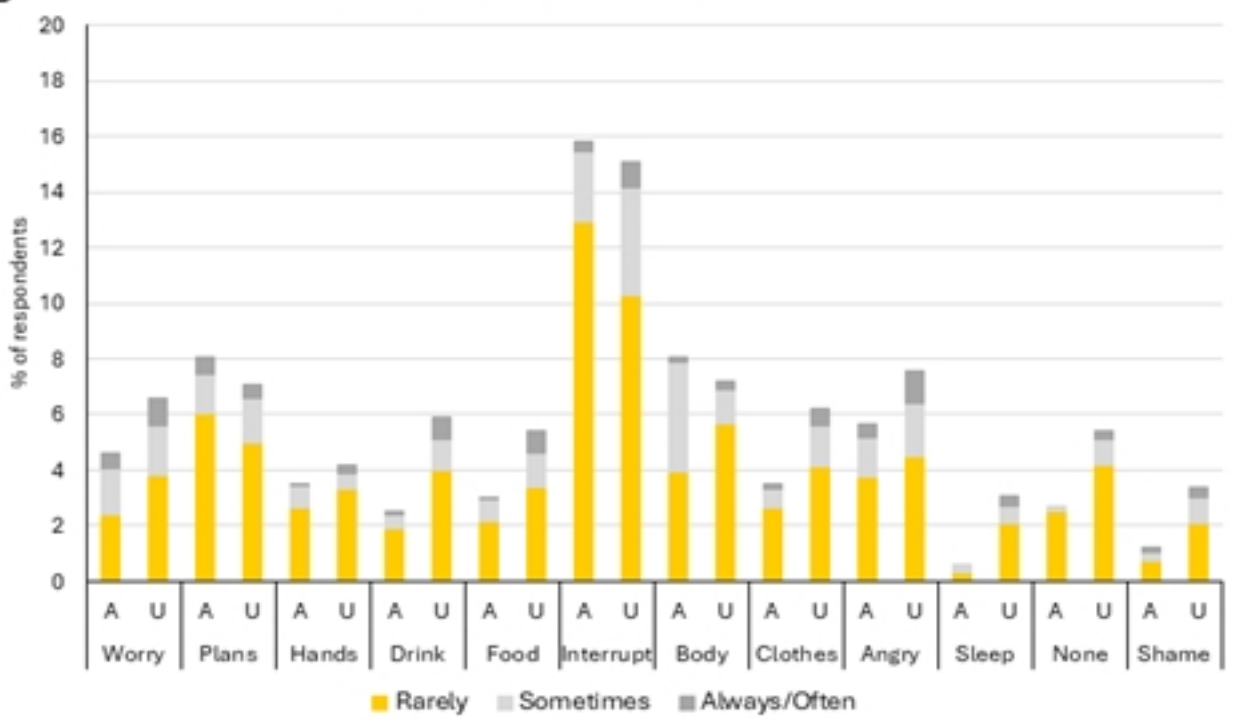
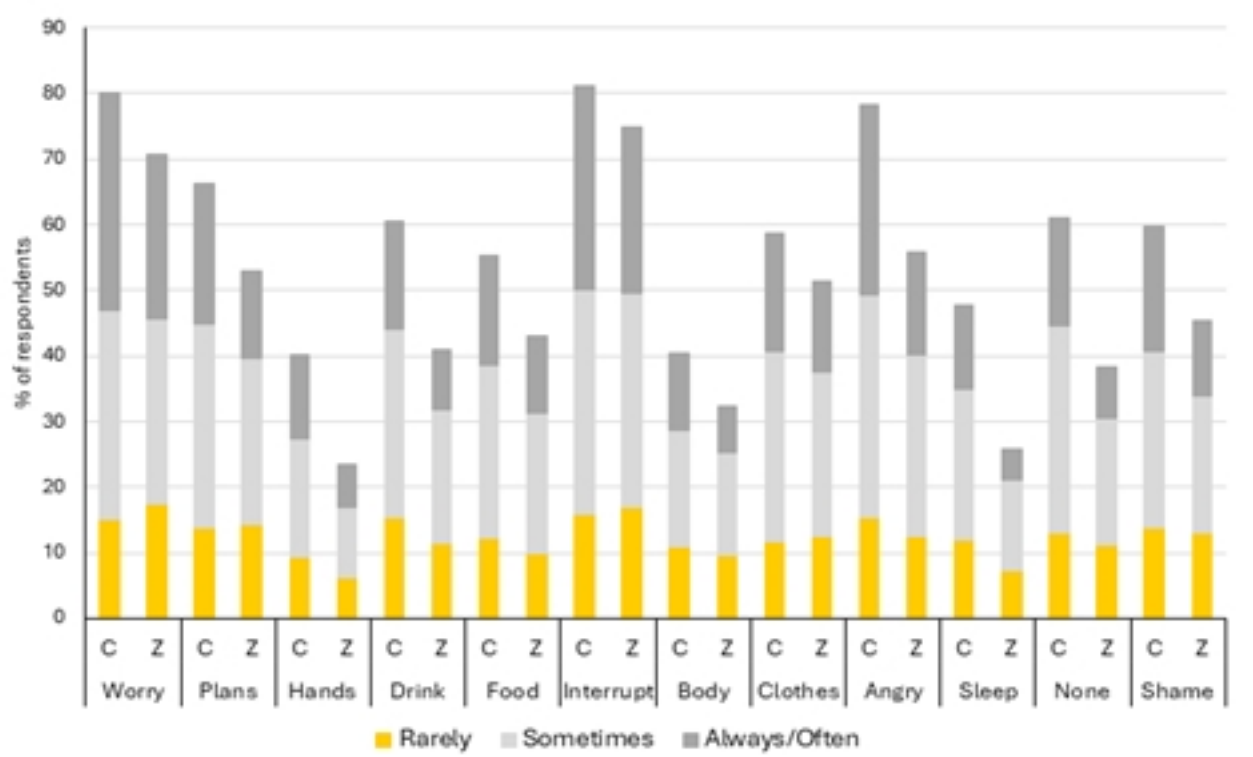
**B****C**

Figure 2

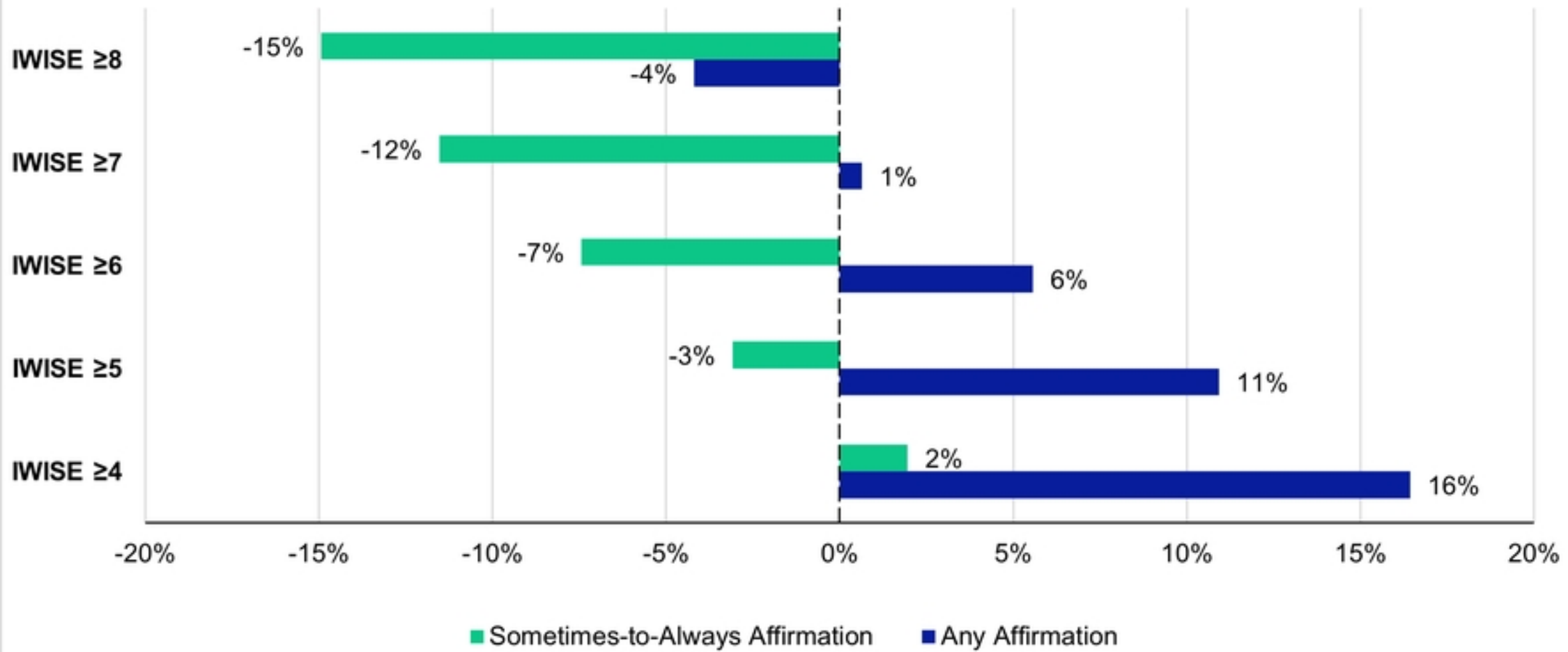


Figure3

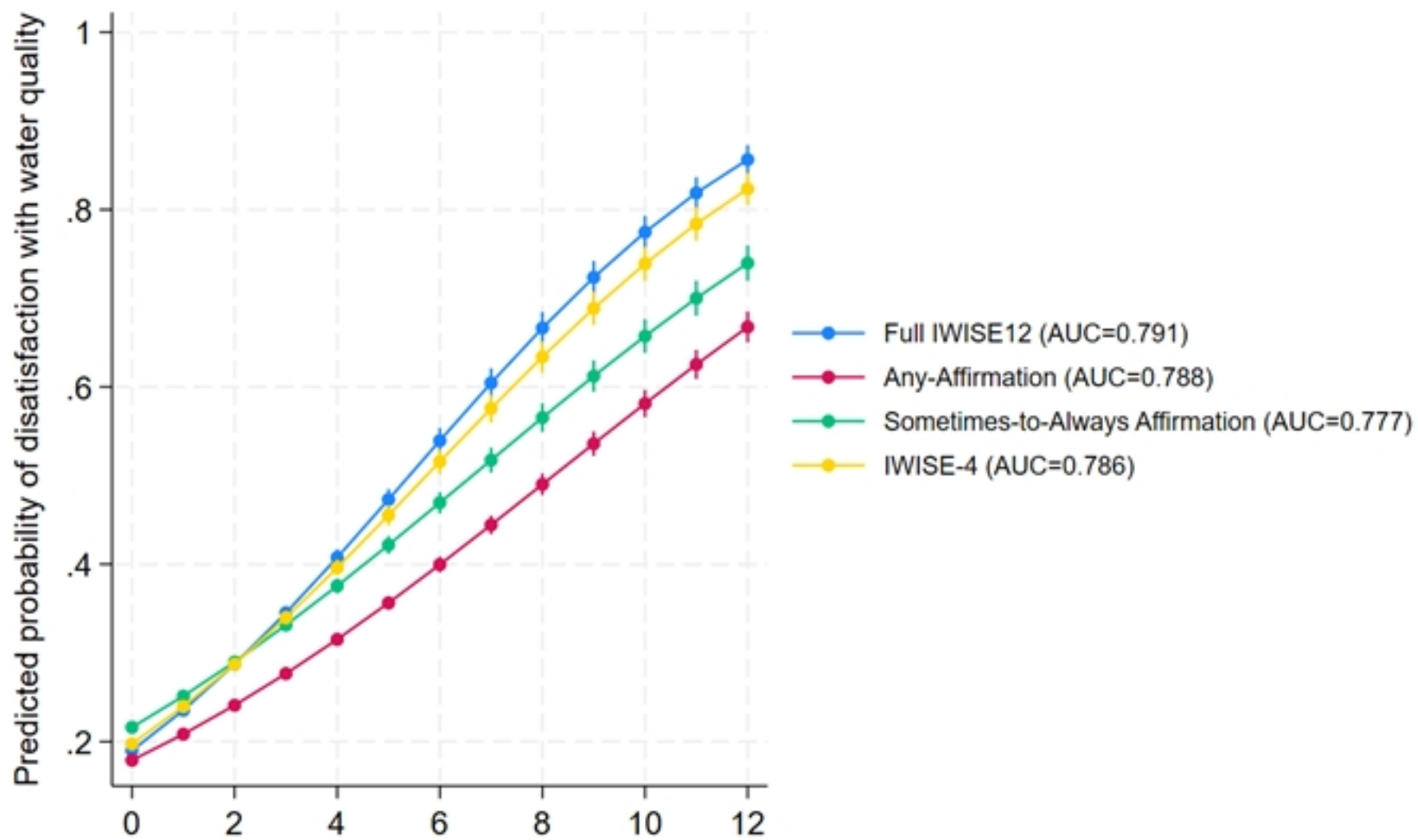


Figure4

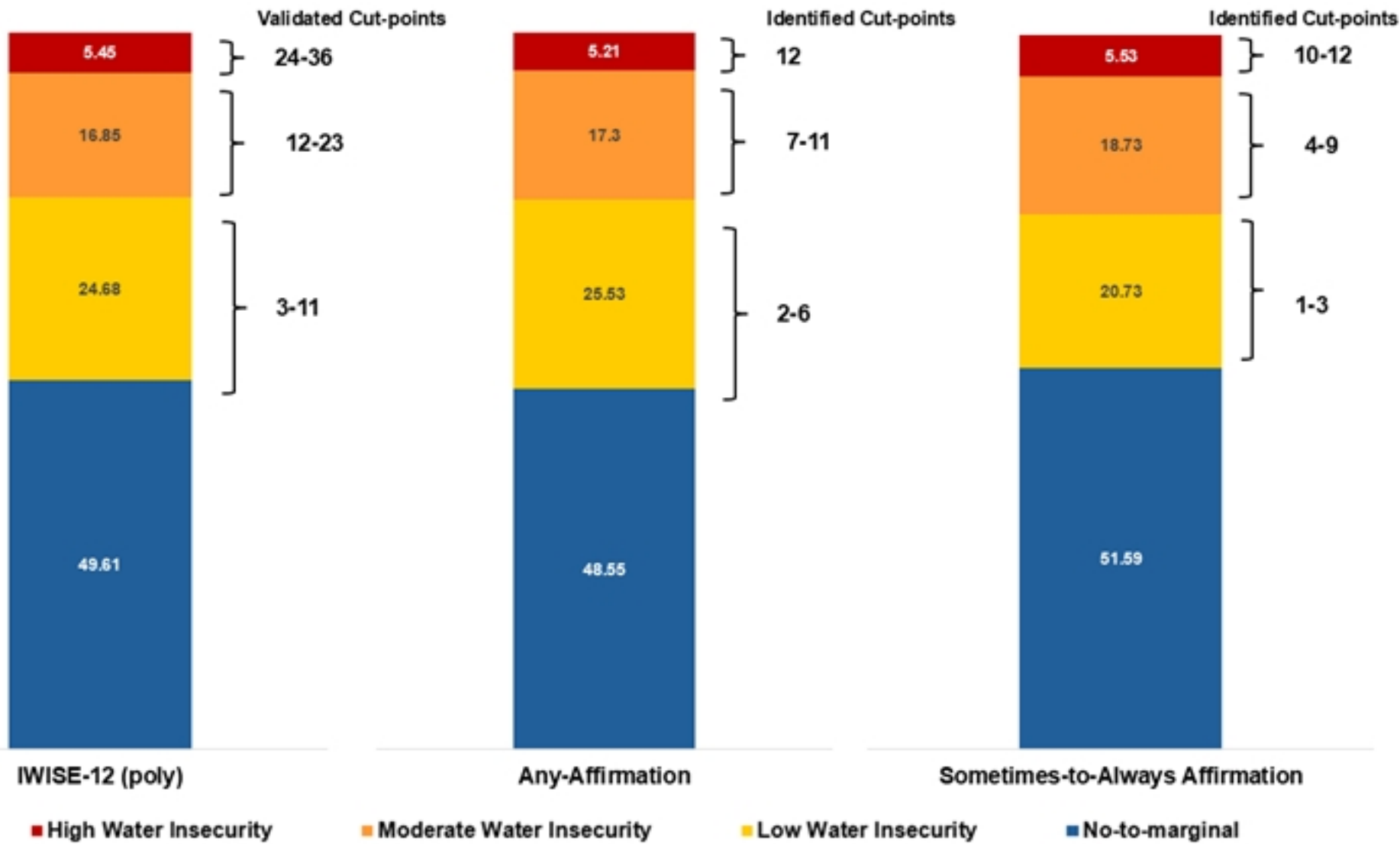


Figure5