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Public Understanding of the Atmospheric River Scale

Zoe N. Caryl,^{a,b} Joseph E. Trujillo-Falcón,^{b, a}, Anna M. Wilson^c, and Abby Bitterman^d

^a Department of Communication, University of Illinois Urbana-Champaign, Urbana, Illinois

*^b Department of Climate, Meteorology & Atmospheric Sciences, University of Illinois Urbana-Champaign,
Urbana, Illinois*

*^c Center for Western Weather and Water Extremes, Scripps Institution of Oceanography, University of
California, San Diego, La Jolla, California*

^d Institute for Public Policy Research and Analysis, University of Oklahoma, Norman, Oklahoma

Corresponding author: Zoe N. Caryl, zoenc2@illinois.edu

17 ABSTRACT

18 Atmospheric rivers (ARs) are long, narrow bands of moisture in the atmosphere that
19 transport large amounts of water vapor, producing hazards ranging from heavy rain to high
20 winds once they reach land. In 2019, researchers at the Center for Western Weather and
21 Water Extremes, with partners including the National Weather Service and the California
22 Department of Water Resources, developed the AR Scale, which ranges from one to five and
23 categorizes the intensity and potential impacts of ARs to improve communication and public
24 understanding of these events. This study presents the first nationwide, bilingual
25 (English/Spanish) quantitative assessment of the AR Scale in the United States. Using a
26 cross-sectional survey, we examine three key areas: (1) perceived benefits and risks of AR
27 events at different levels, (2) thresholds for taking protective actions during an event based on
28 the AR Scale, and (3) category comprehension of hazard potential. Results provide insight
29 into current levels of public understanding and may help improve the design of forecast
30 products and communication tools used during ARs, supporting better decision-making and
31 response during these high-impact weather events.

32 SIGNIFICANCE STATEMENT

33 Atmospheric rivers (ARs) can bring both beneficial rain and dangerous flooding, and as
34 the AR Scale is being adopted globally, effective communication of AR intensity is
35 increasingly important for public safety. This study provides the first nationwide, bilingual
36 assessment of how the U.S. public interprets the AR Scale. Results show that many
37 participants struggle to correctly order categories, understand impacts, and know when to
38 take protective action, with greater uncertainty among Spanish speakers. By identifying
39 where confusion occurs, this study offers practical ways to improve clarity of AR risk
40 communication across language groups. Improving how ARs are communicated can support
41 clearer decision-making and more equitable preparedness as these events become more
42 frequent with a changing climate.

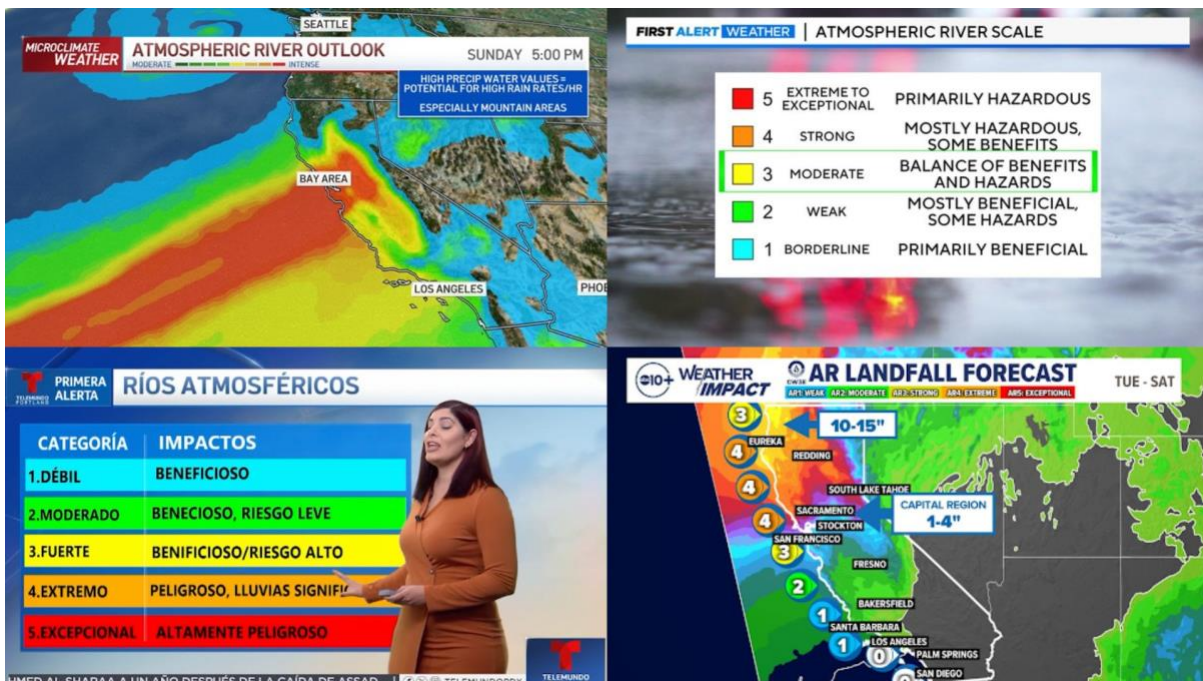
43 CAPSULE

44 A nationwide bilingual survey shows that the public struggles to interpret the
45 Atmospheric River Scale, highlighting key gaps in risk communication and potential
46 opportunities to improve future messaging strategies.

47 **1. Introduction**

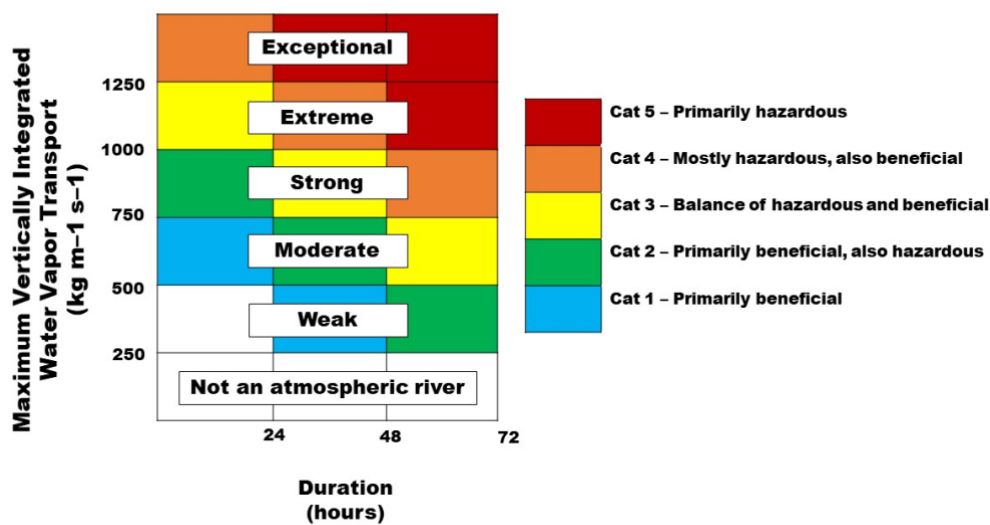
48 Atmospheric rivers (ARs) are long, narrow bands of concentrated moisture in the
49 atmosphere that transport large amounts of water vapor, producing hazards such as flooding
50 and high winds when interacting with terrain (Ralph et al. 2019). A global phenomenon most
51 frequent in mid-latitudes, ARs commonly affect the Western U.S. and other regions across
52 the country, including the Southeast, the central United States, and the Northeast (Mahoney et
53 al. 2016; Moore et al. 2012; Debbage et al. 2017). From 1978 to 2017, ARs drove 84% of
54 major river floods in the Western U.S., with estimated damages totaling \$50.8 billion (Ralph
55 et al. 2006; Neiman et al. 2013; Corringham et al. 2019). As the climate warms, ARs are
56 projected to be wetter, stronger, and longer-lasting across the globe (Payne et al. 2020).

57 To communicate the risks, hazards, and benefits of ARs to the public, Ralph et al. (2019)
58 developed the Atmospheric River Scale (AR Scale), which has since been used worldwide
59 (Guan et al. 2023). Despite its widespread adoption and recent use by broadcast
60 meteorologists in English and Spanish (Fig. 1), existing evaluations of the AR Scale have
61 focused on its hydrological applications, leaving its effectiveness as a public risk
62 communication tool an open empirical question. With ARs projected to intensify under
63 climate change, it is crucial to obtain needed input on whether the AR Scale is not only
64 understood but also provides enough information to prompt protective action for the public.



65
66 **Fig. 1.** Examples of AR Scale usage across English and Spanish broadcast media in the
67 United States

68 The AR Scale distribution is calculated based on vertically integrated vapor transport
 69 (IVT) (see Ralph et al. 2019 for calculations), with intensity represented by the maximum
 70 IVT value at the location where ARs make landfall. The AR Scale follows a 1-5 ranking from
 71 weak to exceptional, represented as “AR 1” to “AR 5”, “Category 1 to Category 5”, or more
 72 recently, “Level 1 to 5.” Visually, this scale uses colors to express intensity with blue
 73 representing weak or Category 1 and red representing exceptional or Category 5 (Fig. 2).
 74 There are two main exceptions to the scale: the maximum category is “AR 5,” regardless of
 75 whether the event is longer than 48 hours, and some ARs can be classified as too “weak” to
 76 be recognized on the scale based on their duration and intensity.



77
 78 **Fig. 2.** The Atmospheric River Scale, as presented in USGS (2021)

79 By using the duration and intensity framework, this scale is unique in that its categories
 80 primarily distinguish between beneficial and hazardous events. As ARs have higher intensity
 81 and rank higher on the scale from “AR 1 to AR 5,” the potential impacts shift from being
 82 primarily beneficial to primarily hazardous (Fig. 1; Ralph et al. 2019). Primarily beneficial
 83 ARs can replenish water sources, whereas primarily hazardous ARs can cause severe
 84 flooding and wind damage (Guan et al. 2023). Broadcast meteorologists have also used this
 85 wording in their reports to distinguish between AR categories and communicate their
 86 associated impacts to the public in both English and Spanish (Fig. 1).

87 In risk communication, there has been a long-standing history of scales used to warn the
 88 public, whether issued in advance for interpretation or after an event to measure its severity.
 89 A few examples of these scales are the Excessive Rainfall Outlook (ERO) for flooding
 90 (Burke et al. 2023), the Saffir-Simpson Scale for hurricane wind speed (NHC 2025), the

91 Storm Prediction Center (SPC) categorical outlook, used to communicate the likelihood of
92 severe thunderstorm hazards (Ernst et al. 2021), and the NorthEast Snowfall Impact Scale
93 (NESIS) for winter weather (NCEI 2025). Most of these scales range from four to six levels
94 and vary in their use of wording, numbering, and coloring, yet most were developed
95 primarily as operational tools for scientists and forecasters rather than as products designed
96 with public comprehension in mind (Ernst et al. 2026).

97 Like the AR Scale, many use a numerical system to convey hazard intensity. The Saffir-
98 Simpson scale ranks hurricane maximum wind speed from Category 1 to Category 5, with
99 Category 3 representing the threshold for storms considered major hurricanes (NHC 2025).
100 NESIS similarly uses a Category 1 to Category 5 framework for snowfall impacts, pairing
101 each level with words ranging from 'Notable' to 'Extreme' (NCEI 2025). One of the oldest
102 scales combines numbers and words to communicate tornado intensity on a 0–5 scale based
103 on wind speed and damage (Fujita 1971; WSEC 2006). Ripberger et al. (2015) were among
104 the first to empirically test whether such impact words effectively communicate risk to the
105 public alongside the numerical scale, a line of inquiry that has since expanded across multiple
106 hazard types (e.g., Krocak et al. 2022; Ernst et al. 2021).

107 Recently, hazard scales have started to be evaluated in different languages. The SPC
108 categorical outlook uses an English and Spanish version of the scale with increasing English
109 risk terminology of “Marginal,” “Slight,” “Enhanced,” “Moderate,” and “High.” Bitterman et
110 al. (2023) found that the direct translation of this scale from *Elevado* (Enhanced) and *Alto*
111 (High) communicates similar values of risk to Spanish speakers. Trujillo-Falcón et al. (2021)
112 developed an alternative Spanish SPC categorical outlook that was more intuitive for Spanish
113 speakers using the words *Minimo*, *Bajo*, *Moderado*, *Alto*, and *Extremo*, with these changes
114 implemented by the SPC and used in the Spanish versions of the categorical outlook
115 (Bitterman et al. 2023).

116 Given the newness of the AR Scale and its application in English and Spanish, it is vital
117 to assess whether the bilingual public can understand and use this scale to make protective
118 decisions. As a key risk communication tool, the AR Scale conveys information about
119 duration, intensity, and impacts, yet its effectiveness in shaping public comprehension and
120 action remains understudied. Understanding how diverse and bilingual audiences interpret
121 this scale is foundational to designing an AR Scale that people can follow and act on. To fill
122 this gap, we address the following research questions:

123 *RQ1: How does the U.S. public interpret the risk communicated by the AR Scale?*
124 *RQ2: How does the U.S. public perceive the AR Scale categories in terms of risk and*
125 *benefit?*
126 *RQ3: When does the U.S. public begin considering taking protective actions based on*
127 *AR Scale categories?*

128 **2. Methods**

129 The study draws from questions from the 2025 Flooding and Society Survey (FL25;
130 Bitterman et al. 2026a) and the 2025 Flooding and Society Spanish Survey (FLS25;
131 Bitterman et al. 2026b), both led by the Institute for Public Policy Research and Analysis
132 (IPPRA) at the University of Oklahoma. IPPRA fielded FL25 July 2–10, 2025, and FLS25
133 July 7–31, 2025, using self-administered online questionnaires. The University of
134 Oklahoma’s Institutional Review Board (IRB #91418) reviewed and approved both survey
135 instruments and protocols. FL25 enrolled 1,223 U.S. adults, and FLS25 enrolled 375
136 Spanish-speaking U.S. adults, all aged 18 and older, residing in the contiguous United States.
137 A national sample was selected, given that AR forecast products, including the AR Scale, are
138 disseminated through national media channels (Fig. 2), and their impacts are increasingly
139 encountered by publics beyond the Western U.S. (e.g., Debbage et al. 2017).¹ We also opted
140 to center our study on both English and Spanish-speaking audiences, as they constitute the
141 two biggest language groups in the United States (Trujillo-Falcón et al. 2021).

142 Verasight, a survey panel company, recruited participants through multiple methods,
143 including address-based sampling, person-to-person text messaging, and dynamic online
144 targeting. All panel members completed multistep identity verification, including SMS
145 confirmation and Google reCAPTCHA v3, to ensure respondent validity and prevent
146 automated responses. During data collection, Verasight applied a dynamic sampling
147 procedure that adjusted invitation targets in real time to reduce over- or underrepresentation
148 of key demographic groups. Participants completed FL25 in a median of 24.1 minutes and
149 FLS25 in a median of 26.7 minutes.

150 Both surveys targeted the U.S. adult population using benchmarks derived from the 2024
151 American Community Survey (ACS) 1-year Public Use Microdata Sample accessed through

¹ The authors acknowledge that Western U.S. respondents may have greater direct exposure to AR events and familiarity with AR forecast products compared to the broader national sample. However, given the relatively small subsample size for Western U.S. respondents, these results are presented separately rather than alongside the main findings. Analyses of the subset of FL25 and FLS25 respondents residing in the Western U.S., identified through participant-reported ZIP codes, are therefore presented in Appendix A.

152 IPUMS USA (Ruggles et al. 2025). IPPRA constructed post-stratification weights using
 153 iterative proportional fitting (raking), which iteratively adjusts respondent weights until the
 154 marginal distributions of key demographic variables in the weighted sample converge to
 155 population benchmarks. For FL25, the weighting procedure incorporated age, gender, race,
 156 ethnicity, education, income, and NWS region. For FLS25, it incorporated age, gender,
 157 Hispanic heritage, English language proficiency, education, income, and NWS (n.d.) region,
 158 as previous work in bilingual risk communication has established these factors as important
 159 benchmarks (Trujillo-Falcón 2026). Table 1 presents the demographic composition of both
 160 survey samples alongside population benchmarks.

	FL25 (English, n = 1,223)		FLS25 (Spanish, n = 375)	
	U.S. Adult Population (%)	Participants (%)	Spanish-Speaking U.S. Adult Population (%)	Participants (%)
Gender				
Female	51.0	55.8	50.0	74.4
Male	49.0	44.2	50.0	25.6
Age				
18–29	20.1	16.1	24.2	14.7
30–49	33.8	34.7	40.0	56.3
50–64	23.2	25.3	22.3	26.1
65 and up	22.9	23.9	13.5	2.9
Education				
High School or Less	37.3	30.7	56.9	34.9
Some College/2-year Degree	28.5	30.1	24.1	39.5
4-year/Post-Graduate Degree	34.2	39.2	19.0	25.6
Race				
White	59.2	60.7	N/A	N/A
Black or African American	11.6	11.9	N/A	N/A
Other race	11.0	9.7	N/A	N/A

Ethnicity				
Hispanic	18.2	17.7	N/A	N/A
Non-Hispanic	81.8	82.3	N/A	N/A
Hispanic Heritage				
Not Hispanic	N/A	N/A	5.4	2.9
Mexican	N/A	N/A	52.6	17.9
Other Hispanic	N/A	N/A	42.0	79.2
English Language Proficiency				
Does not speak English	N/A	N/A	10.3	12.3
Yes, but not well	N/A	N/A	16.4	44.3
Yes, speaks well	N/A	N/A	18.2	26.1
Yes, speaks very well	N/A	N/A	55.2	17.3
Income				
Less than \$50,000	25.9	29.8	29.9	71.2
\$50,000 or more	74.1	70.2	70.1	28.8
NWS Region				
Eastern	31.4	30.8	21.1	25.3
Southern	27.8	28.8	36.7	53.1
Central	20.3	19.1	10.0	7.7
Western	20.5	21.3	32.2	13.9

161 **Table 1.** Demographic composition of FL25 (English) and FLS25 (Spanish) survey
162 samples compared to U.S. adult population benchmarks.

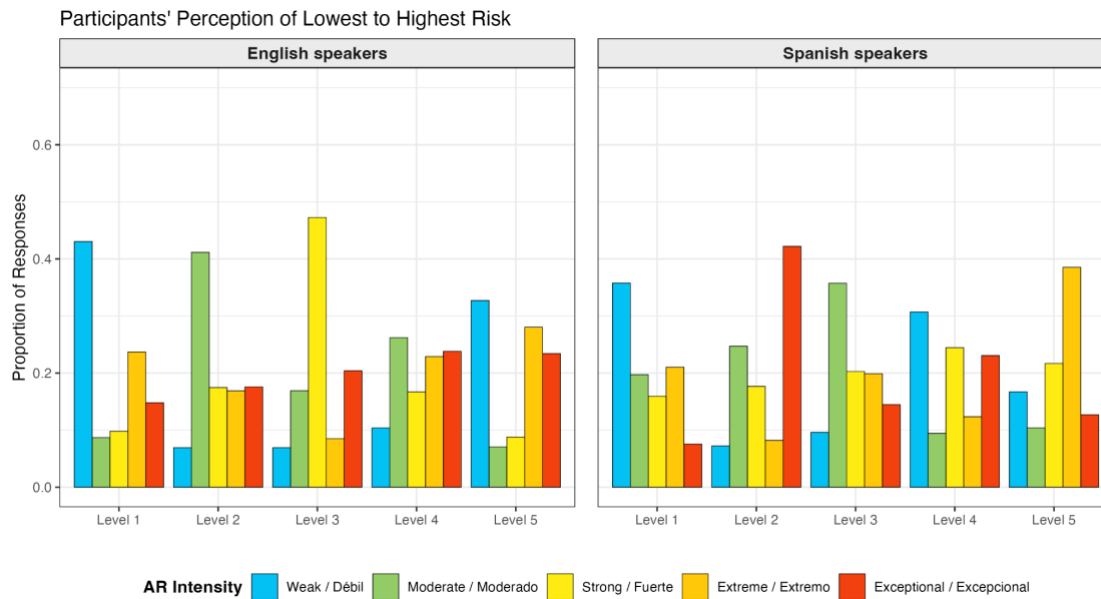
163 Our study examined three sets of AR-specific items embedded in both surveys (for exact
164 question wording, visit Bitterman et al. 2026 a,b). To assess perceived ordering of AR Scale
165 intensity, participants ranked five AR Scale terms– “Weak,” “Moderate,” “Strong,”
166 “Extreme,” and “Exceptional” (or *Débil*, *Moderado*, *Fuerte*, *Extremo*, and *Excepcional* in
167 Spanish)– from 1 (lowest risk) to 5 (highest risk) based on how serious each term sounds,
168 following previous work that has examined categorical risk terminology (Bitterman et al.
169 2023; Ernst et al. 2021). To assess perceived impact, participants characterized the expected
170 impacts of each AR category (AR1–AR5) as primarily beneficial (*primordialmente*
171 *beneficioso*), equally beneficial and hazardous (*igualmente beneficioso como peligroso*),

172 primarily hazardous (*primordialmente peligroso*), or not sure, following the original
173 definitions of the AR Scale (Ralph et al. 2019). To assess perceived protective action
174 thresholds, participants were asked to indicate at which AR category they would begin to
175 consider each of seven actions in a hypothetical AR event scenario: (1) seeking additional
176 information, (2) preparing emergency supplies, (3) considering protective action, (4) securing
177 outdoor items, (5) reconsidering travel plans, (6) preparing for power outages, and (7)
178 evacuating. Each action also included a “would not take this action in any scenario option.”

179 We conducted all analyses in R version 2026.01.2+418 using the *srvyr* package
180 (Freedman et al 2026) to incorporate post-stratification weights. For each survey item, we
181 calculated weighted proportions and 95% confidence intervals. For the scale ordering task,
182 we additionally calculated weighted mean perceived ranks for each AR intensity term and
183 identified the most common response patterns across all five rankings. This analytical
184 approach is consistent with prior weather risk communication studies that have used
185 representative surveys to evaluate public interpretation of hazard scales and provided
186 recommendations for NWS agencies (Bitterman et al. 2023; Ernst et al. 2021). Results from
187 FL25 (English) and FL25 (Spanish) are presented in two-panel figures to facilitate direct
188 comparison between language groups.

189 **3. Results**

190 Fig. 3 presents the weighted distributions of English and Spanish speakers’ rankings of
191 each AR intensity term across Levels 1–5. Among English speakers, “Weak” accounted for
192 the largest share of responses at Level 1 (43.0%) and “Strong” at Level 3 (47.3%). However,
193 participants showed considerable variation at the upper end of the scale. At Level 4,
194 “Moderate” received the highest proportion of responses (26.2%), followed closely by
195 “Exceptional” (23.8%) and “Extreme” (22.9%). At Level 5, “Weak” received the largest
196 share of responses (32.7%), with Extreme (28.0%) and Exceptional (23.4%) following.
197 Spanish speakers showed a comparable pattern for *Débil* at Level 1 (35.8%), but considerably
198 greater dispersion across the middle and upper categories. *Excepcional* received its highest
199 proportion at Level 2 (42.2%) rather than Level 5, and at Level 5, *Extremo* was most
200 frequently selected (38.5%) while *Excepcional* was assigned by only 12.7% of participants.

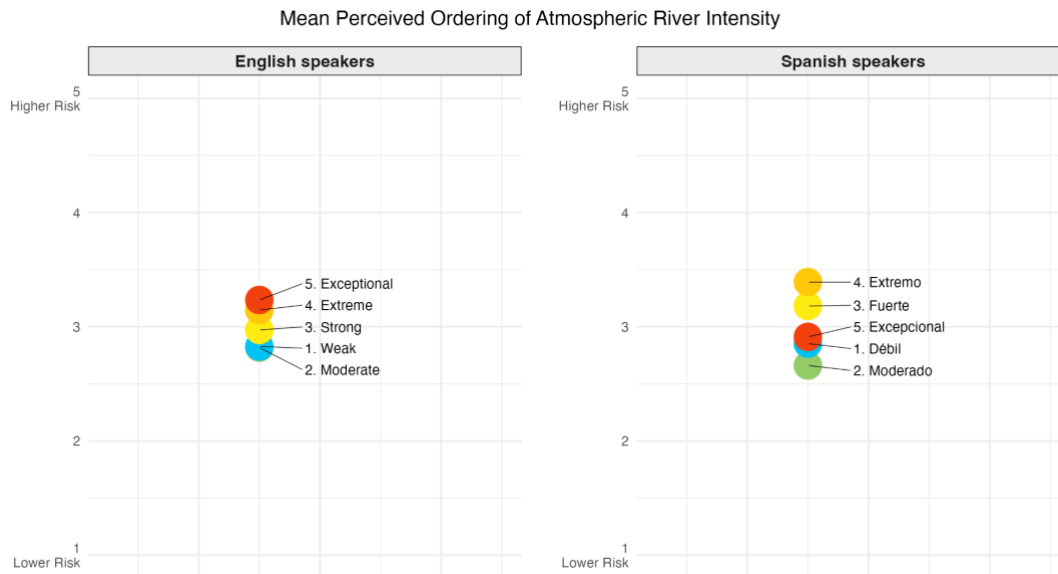


201

202 **Fig. 3.** Distributions of English and Spanish survey participants' rankings of the AR Scale
 203 categorical words.

204 The share of English speakers assigning “Weak” to Level 5 (32.7%) warranted additional
 205 investigation. To rule out data quality issues, we conducted three robustness checks: (1)
 206 rerunning the analysis with unweighted data, (2) removing respondents who spent fewer than
 207 5 seconds on the ranking questions, and (3) removing respondents who spent fewer than 10
 208 seconds on the ranking questions. All three approaches produced results substantively
 209 identical to those reported in Fig. 3, suggesting the pattern reflects genuine participant
 210 responses rather than inattentive or random answering.

211 Fig. 4 plots the weighted mean perceived rank for each intensity term. Among English
 212 speakers, all five terms fell between mean ranks of 2.82 (“Moderate”) and 3.23
 213 (“Exceptional”), a range of only 0.41 points on a 1–5 scale. The intended ordering was
 214 partially preserved, with “Moderate” (2.82) and “Weak” (2.83) receiving the lowest mean
 215 ranks and “Strong” (2.97), “Extreme” (3.15), and “Exceptional” (3.23) receiving
 216 progressively higher ranks. Among Spanish speakers, all five terms similarly clustered near
 217 the midpoint, ranging from 2.66 (*Moderado*) to 3.39 (*Extremo*), a range of 0.73 points.
 218 However, the intended ordering was not preserved: *Excepcional* (2.91) received a lower mean
 219 rank than both *Fuerte* (3.18) and *Extremo* (3.39), placing the highest intensity term third
 220 rather than fifth. *Débil* (2.85) received the second lowest mean rank, broadly consistent with
 221 its intended position, but *Moderado* (2.66) received the lowest mean rank overall.



222

223 **Fig. 4.** Mean placement of each AR Scale term among English and Spanish participants

224 Table 2 presents the five most common ordering patterns among English and Spanish
 225 speakers. Among English speakers, the most common response was “Weak” →
 226 “Moderate” → “Strong” → “Exceptional” → “Extreme” (27.78%), which correctly places
 227 the lower three categories but inverts “Extreme” and “Exceptional” at the upper end. The
 228 official ordering (i.e., “Weak” → “Moderate” → “Strong” → “Extreme” → Exceptional) was
 229 the second most common response (26.51%), and the two most prevalent patterns
 230 differed only in the placement of the two highest-intensity terms. The third and fourth
 231 most common patterns (12.70% and 11.11%, respectively) represented complete reversals
 232 of the scale from highest to lowest risk.

Top Five Order of Common Responses (English)	n	%
Weak → Moderate → Strong → Exceptional → Extreme	175	27.78
Weak → Moderate → Strong → Extreme → Exceptional	167	26.51
Extreme → Exceptional → Strong → Moderate → Weak	80	12.70
Exceptional → Extreme → Strong → Moderate → Weak	70	11.11
Extreme → Strong → Exceptional → Moderate → Weak	69	10.95

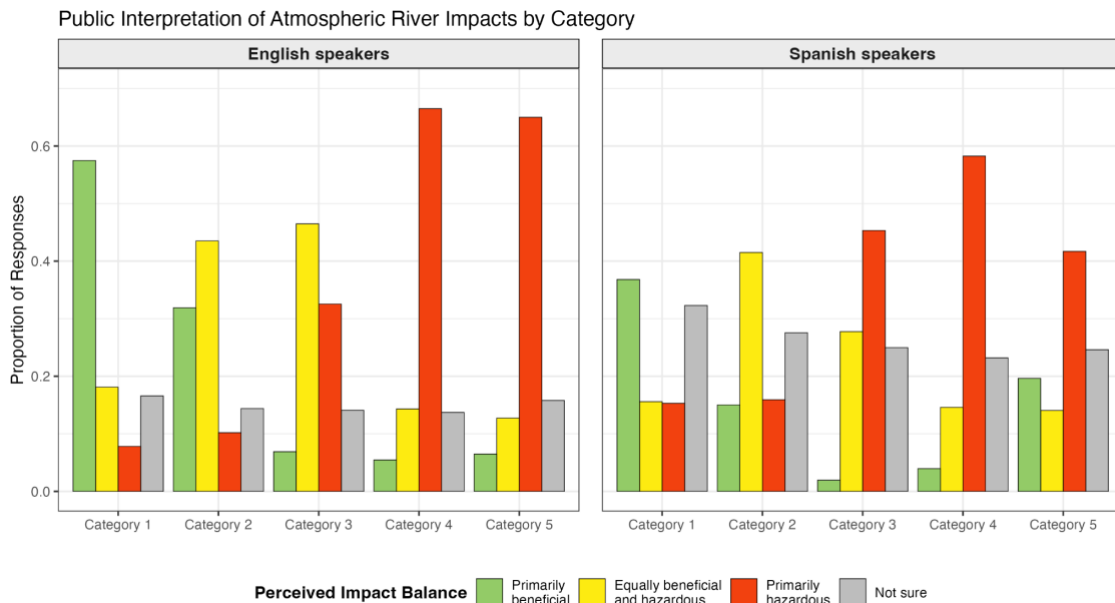
233 **Table 2.** Top five most common ordering patterns of AR Scale intensity terms among
 234 English-speaking survey respondents (n = 1,223), with the official AR Scale ordering
 235 indicated in **red font**.

236 Among Spanish speakers (Table 3), the correct ordering (*Débil* → *Moderado* → *Fuerte* →
 237 *Extremo* → *Excepcional*) was most frequently selected (27.27%), with the second most
 238 common pattern transposing *Excepcional* and *Extremo* (26.36%). As with English speakers,
 239 the two most prevalent patterns were nearly equivalent in frequency and differed only in the
 240 placement of the two highest-intensity terms.

Top Five Order of Common Responses (Spanish)	n	%
Débil → Moderado → Fuerte → Extremo → Excepcional	30	27.27
Débil → Moderado → Excepcional → Fuerte → Extremo	29	26.36
Débil → Moderado → Fuerte → Excepcional → Extremo	13	11.82
Débil → Excepcional → Moderado → Fuerte → Extremo	10	9.09
Débil → Extremo → Moderado → Fuerte → Excepcional	7	6.36

241 **Table 3.** Top five most common ordering patterns of AR Scale intensity terms among
 242 English-speaking survey respondents (n = 375), with the official AR Scale ordering indicated
 243 in **red font**.

244 Fig. 5 presents the weighted distributions of English and Spanish speakers’
 245 characterizations of expected impacts for each AR category. Among English speakers, a
 246 majority identified AR1 (“Weak”) as primarily beneficial (57.5%), and AR4 (“Extreme”) and
 247 AR5 (“Exceptional”) were most frequently characterized as primarily hazardous (66.5% and
 248 65.0%, respectively). AR2 (“Moderate”) was most commonly characterized as equally
 249 beneficial and hazardous (43.5%), and AR3 (“Strong”) followed a similar pattern (46.5%).



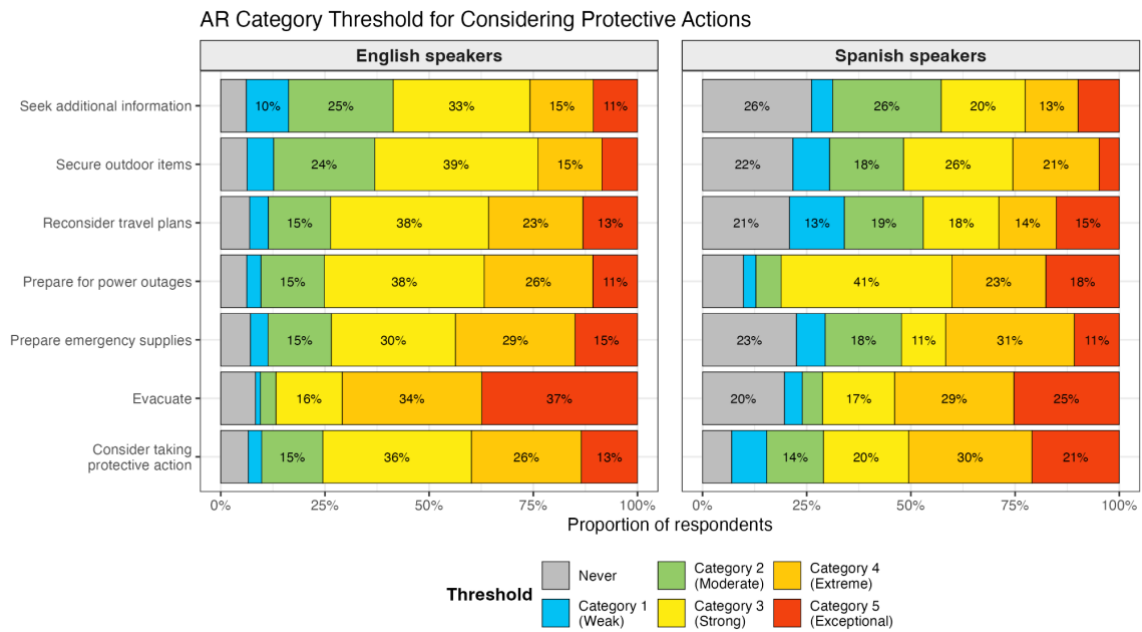
250

251 **Fig. 5.** Distributions of English and Spanish participants' characterizations of expected
 252 impacts by AR category.

253 Spanish speakers exhibited a similar pattern, with AR3 (*Fuerte*) through AR5
 254 (*Excepcional*) most frequently characterized as primarily hazardous (45.3%, 58.2%, and
 255 41.7%, respectively), but with three notable differences. First, "not sure" responses were
 256 substantially higher across all five categories among Spanish speakers, ranging from 23.2%
 257 to 32.3%, compared to 13.7% to 16.6% among English speakers. Second, Spanish speakers
 258 were less likely to characterize AR5 (*Excepcional*) as primarily hazardous (41.7%) compared
 259 to English speakers (65.0%), and 19.6% of Spanish speakers characterized AR5 as primarily
 260 beneficial, compared to 6.5% of English speakers. Third, AR1 (*Débil*) elicited a lower rate of
 261 "primarily beneficial" responses among Spanish speakers (36.8%) compared to English
 262 speakers (57.5%), accompanied by a higher "not sure" rate (32.3% vs. 16.6%).

263 Fig. 6 presents the AR category at which English and Spanish speakers reported they
 264 would begin to consider each of seven protective actions in a hypothetical AR event scenario.
 265 Among English speakers, the largest share of respondents indicated they would seek
 266 additional information beginning at AR3 (32.8%), followed by AR2 (25.1%). Stated
 267 thresholds for securing outdoor items (AR3: 39.1%), reconsidering travel plans (AR3:
 268 37.9%), and preparing for power outages (AR3: 38.3%) followed comparable patterns, with
 269 AR3 as the modal threshold for each action. Stated thresholds for preparing emergency
 270 supplies were more evenly distributed across AR3 (29.8%) and AR4 (28.7%). Evacuation

271 exhibited the highest stated threshold, with the largest shares of English speakers citing AR4
 272 (33.5%) or AR5 (37.4%), and 8.3% reporting they would not evacuate under any scenario.



273
 274 **Fig. 6.** Stated protective action thresholds by AR category among English and Spanish
 275 participants.

276 Spanish speakers exhibited broadly similar distributions but with two notable differences.
 277 First, "would not take this action in any scenario" responses were higher among Spanish
 278 speakers across several actions, most notably for seeking additional information (26.1% vs.
 279 6.1%), securing outdoor items (21.7% vs. 6.3%), and reconsidering travel plans (20.9% vs.
 280 6.9%). Second, a higher percentage of Spanish speakers reported they would not evacuate
 281 under any scenario (19.7% vs. 8.3% among English speakers), though among those who
 282 indicated a threshold, responses were broadly comparable at the higher categories, with
 283 28.6% citing AR4 and 25.3% citing AR5. Given the smaller Spanish sample (n = 375),
 284 confidence intervals for Spanish estimates are wider throughout, and these results should be
 285 interpreted accordingly.

286 **4. Discussion**

287 When Ralph et al. (2019) introduced the AR Scale, they provided forecasters and
 288 communicators with a reliable framework for conveying AR intensity and impact to the
 289 public. However, whether the public understands that framework it is meant to serve remains

290 an open empirical question. This study presents the first nationwide bilingual assessment of
291 public understanding of the AR Scale, finding that comprehension of the scale's intended
292 intensity gradient is incomplete among both English and Spanish speakers, with notable
293 differences between language groups across all three dimensions examined: intensity
294 ordering, perceived benefits and hazards, and protective action thresholds.

295 **a. AR Scale Ordering.** Categorical risk scales depend on language that intuitively
296 conveys increasing levels of risk to the public, yet the words chosen to represent scale
297 categories across various weather hazards have frequently proven difficult for the public to
298 interpret accurately (Bitterman et al. 2023; Ernst et al. 2021; Krocak et al. 2022; Trujillo-
299 Falcón et al. 2021; Williams et al. 2022). When applied to the AR Scale, both English and
300 Spanish speakers showed difficulty ordering its intensity terms. The narrow spread of the
301 weighted mean ranks—less than half a point across all five terms on a five-point scale (Fig.
302 2)—suggests that the current AR terms do not communicate a clearly differentiated gradient of
303 risk, a finding with implications for how the scale is designed and presented.

304 At the upper end of the scale, two terms appear to be driving much of the observed
305 difficulty. "Exceptional" and *Excepcional* can carry positive connotations in everyday usage
306 in both languages, which may have led some participants to associate the term with lower risk
307 rather than the scale's highest category. Among English speakers, a substantial share also
308 placed "Weak" in the highest-risk position, a pattern confirmed in multiple robustness checks
309 and in the Western U.S. subsample (Fig. A1). One possible explanation for this
310 counterintuitive placement may relate to the recognition of compound AR sequences, where
311 weaker events following stronger ones can still produce extreme impacts (Zhou et al. 2024).
312 Future work should examine whether prior exposure to compound AR sequences shapes
313 public risk perception of lower-category events through longitudinal designs, as has been
314 done with hurricanes (Demuth et al. 2025).

315 Among Spanish speakers, the difficulty at the upper end of the scale is further consistent
316 with Bitterman et al. (2023), who found that *Extremo* has been among the highest risk terms
317 tested in Spanish-language hazard communication. The SPC categorical outlook may further
318 reinforce this pattern: the first hazard scale to debut nationwide in Spanish uses *Extremo* as
319 its highest category (Trujillo-Falcón et al. 2021). Future work should more deeply examine
320 Spanish-speaking communities as a heterogeneous group, exploring how factors such as
321 dialect group, English-language proficiency, and foreign-born status shape interpretation of

322 the AR Scale, as previous work has documented differences in risk perceptions across these
323 groups (Trujillo-Falcón et al. 2022).

324 ***b. Perceived Benefits and Hazards.*** Public characterizations of AR impacts broadly
325 aligned with the beneficial-to-hazardous gradient embedded in the AR Scale's design. English
326 speakers correctly identified AR1 as primarily beneficial and AR4–AR5 as primarily
327 hazardous at high rates, indicating that the scale's core risk communication logic (i.e., that
328 higher categories signal greater hazard) is at least partially accessible to English-speaking
329 audiences. AR3, designated as the balance point between beneficial and hazardous impacts in
330 the scale's original framework (Ralph et al. 2019), was most commonly characterized as
331 equally beneficial and hazardous, suggesting reasonable alignment at the midpoint. However,
332 AR2, which the scale designates as mostly beneficial, was also frequently characterized as
333 equally beneficial and hazardous, suggesting the distinction between the two lowest
334 categories may not be clearly communicated by the scale's current design. Future work
335 should examine whether supplementary impact language or visual design changes could
336 better differentiate AR2 from AR3, particularly given the practical implications of
337 mischaracterizing a mostly beneficial event as equally hazardous.

338 Spanish speakers exhibited a broadly similar beneficial-to-hazardous gradient, but with
339 patterns that warrant closer attention. The substantially higher “not sure” rates across all five
340 categories likely reflect lower familiarity with ARs specifically, consistent with previous
341 work showing that while weather salience is generally high among Spanish-speaking
342 communities, awareness of specific hazard types may be limited by differences in disaster
343 subculture (Gaviria Pabón et al. 2025; Trujillo-Falcón et al. 2024a). Most notably, a
344 substantial share of Spanish speakers characterized AR5 as primarily beneficial, a finding
345 with direct safety implications given that AR5 represents the scale's most dangerous category
346 and is associated with the greatest potential for flooding and loss of life. Future work should
347 examine whether targeted AR education and Spanish-language outreach can improve
348 comprehension of the upper categories, particularly for communities in high-risk regions.

349 ***c. Protective Action Thresholds.***

350 Stated thresholds for most protective actions clustered around AR3 among English
351 speakers, with evacuation concentrating at AR4 and AR5. The concentration of information-
352 seeking and preparatory action thresholds at AR3 is noteworthy given that AR3 represents
353 the threshold at which the AR Scale's own framework begins to characterize events as

354 primarily hazardous (Ralph et al. 2019), suggesting some intuitive alignment between scale
355 design and public stated thresholds among English speakers. Evacuation, the highest-stakes
356 action examined, followed patterns consistent with prior work finding that publics report
357 higher stated thresholds for more disruptive protective actions (Morss et al. 2016; Kang et al.
358 2007).

359 Spanish speakers showed broadly comparable threshold distributions among those who
360 reported a threshold, but differed substantially in the proportion reporting they would not take
361 a given action under any scenario. Elevated 'would not act' rates for information seeking,
362 securing outdoor items, and reconsidering travel, as well as a substantially higher rate of
363 reported non-evacuation, should not be interpreted solely as reduced AR Scale
364 comprehension. Prior research has documented that non-action in Spanish-speaking and
365 immigrant communities is often shaped by structural constraints, limited access to Spanish-
366 language hazard information, and reduced institutional trust (Trujillo-Falcon et al. 2024b).
367 Future work should examine whether these patterns reflect perceived inapplicability of the
368 hypothetical scenario, distrust of official messaging, or material constraints on action.

369 **d. Limitations.** Several limitations of this study warrant consideration. First, the AR Scale
370 includes components beyond wording, such as colors, which could be tested in the future, as
371 in Ernst et al. (2021). Second, the survey did not test public understanding of the *duration*
372 component of the AR Scale, which Ralph et al. (2019) built in as a key parameter. Third,
373 while a nationwide sample is appropriate given the AR Scale's increasing use beyond the
374 Western U.S., regional samples in high-exposure areas would allow for more targeted
375 recommendations for communities with direct AR experience, beyond the exploratory
376 Western U.S. subsample presented in Appendix A. Fourth, the Spanish-speaking sample was
377 smaller than the English-speaking sample, resulting in wider confidence intervals throughout,
378 and should be interpreted accordingly. Finally, all protective action items were presented as
379 hypothetical scenarios, and stated thresholds may not correspond to actions taken during
380 actual AR events, a limitation shared by survey-based hazard communication research more
381 broadly (Morss et al. 2016). Future work could examine protective action decisions
382 longitudinally during actual AR events, pairing survey data with real-time hazard information
383 to better understand how stated thresholds translate into behavior under realistic conditions

384 **5. Advancing the AR Scale for the U.S. Public**

385 The AR Scale represents an important advance in communicating atmospheric river
386 intensity and potential impacts to the public (Ralph et al. 2019), and the findings presented
387 here offer an empirical foundation for its continued development as a risk communication
388 tool. Several priorities emerge for advancing the scale across diverse populations.

389 Based on the findings presented here, several practical recommendations are offered for
390 NWS forecasters, broadcast meteorologists, and emergency managers in advance of more
391 systematic testing of alternative scale terminology. First, current AR Scale products should
392 consider supplementing intensity terms with explicit impact language across all five
393 categories to enhance understanding. This approach is consistent with the NWS impact-based
394 warning framework, where the addition of specific hazard and impact language has been
395 shown to improve risk perception and promote protective action (Casteel 2016). Second,
396 future scale design efforts should examine whether numerical labels alone communicate risk
397 more clearly. Krocak et al. (2022) found that numeric level information reduced misordering
398 of the SPC categorical scale relative to verbal labels alone, suggesting a similar approach
399 may improve public interpretation of the AR categories. Third, Spanish-language AR Scale
400 terminology should be reviewed for consistency with existing bilingual hazard scales,
401 particularly given the SPC categorical outlook's established use of *Extremo* as its highest
402 category (Bitterman et al. 2023). Finally, while media training efforts such as those led by the
403 Scripps Institution of Oceanography (n.d.) have begun to equip broadcast meteorologists with
404 AR Scale communication tools, future outreach should extend beyond media professionals
405 toward broader public educational campaigns, particularly for Spanish-speaking
406 communities, where structural barriers to protective action were most evident.

407 Addressing the terminology challenges identified in our study will require systematic
408 testing of alternative intensity descriptions, particularly at the upper end of the scale where
409 AR4 and AR5 were poorly differentiated by both language groups. Prior work in hazard
410 communication has shown that small changes in terminology can alter perceived severity and
411 stated protective action (Trujillo-Falcón et al. 2021; Bitterman et al. 2023). Importantly, the
412 Spanish-language AR Scale should not be treated as a direct translation of the English version
413 but rather evaluated as an independent system shaped by distinct linguistic and cultural
414 interpretations of the terminology (Bitterman et al. 2023; Trujillo-Falcón et al. 2022). The
415 elevated 'would not act' rates among Spanish speakers further highlight the need to
416 distinguish between perceived understanding of risk and perceived ability to respond, which
417 are fundamentally different dimensions of protective decision-making.

418 Beyond the AR Scale itself, situating it within the broader ecosystem of hazard
419 communication products is an important next step. The Excessive Rainfall Outlook (Burke et
420 al. 2023), SPC categorical outlooks, and WSSI may all operate simultaneously during high-
421 impact AR events, yet their combined influence on public interpretation remains poorly
422 understood. Comparative research may help determine whether consistency across scales
423 improves comprehension or whether variation introduces uncertainty in how risk is perceived
424 (Ernst et al. 2021).

425 Complementing these quantitative directions, qualitative work with both English and
426 Spanish-speaking populations would offer insights that surveys alone cannot provide.
427 Interviews and focus groups may help clarify how individuals interpret AR terminology in
428 context, understand overlapping hazard products, and how these interpretations are shaped by
429 lived experience and structural constraints. Integrating qualitative insights with survey-based
430 findings may help provide a more complete foundation for refining the AR scale as both a
431 meteorological and communicational tool. As the AR Scale continues to evolve as a
432 communicative product, its success will ultimately depend on how well it supports a timely
433 and equitable response across *all* communities.

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441 *Data Availability Statement.*

442 The datasets analyzed in this study are available in the Harvard Dataverse. English-
443 language data are described in Bitterman et al. (2026a) or at
444 <https://doi.org/10.7910/DVN/5VYPUQ>. Spanish-language data are described in Bitterman et
445 al. (2026b) or at <https://doi.org/10.7910/DVN/QGDEKA>.

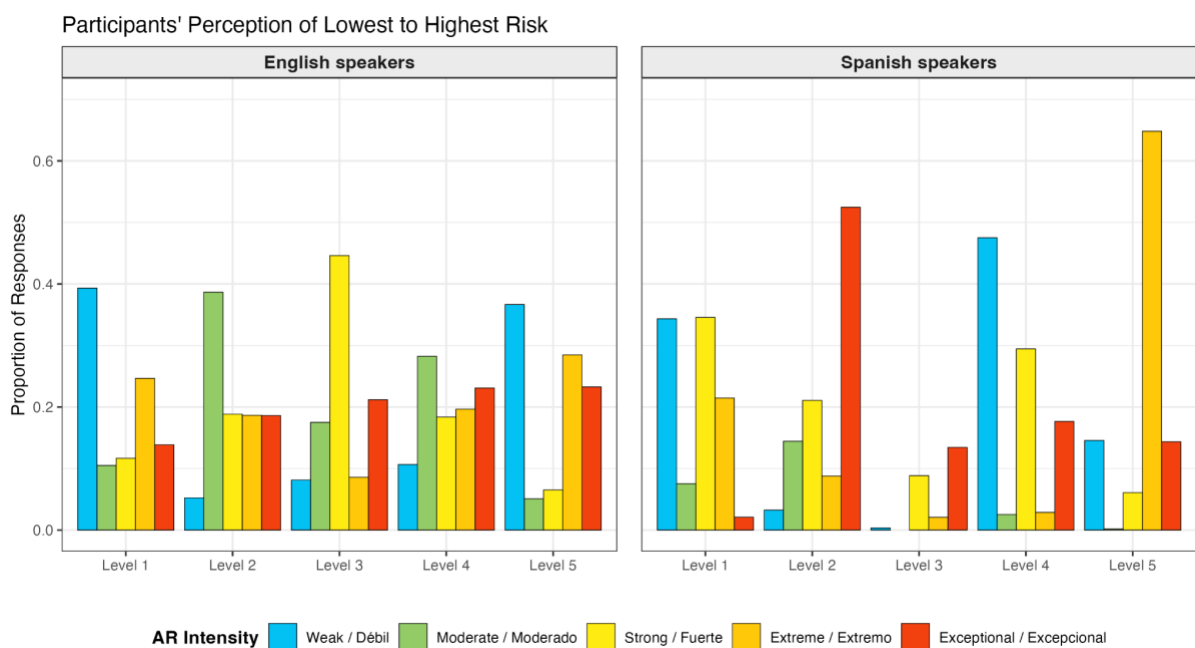
446 APPENDIX

447 **Appendix A**

448 *An Analysis of the Western U.S. Subsample*

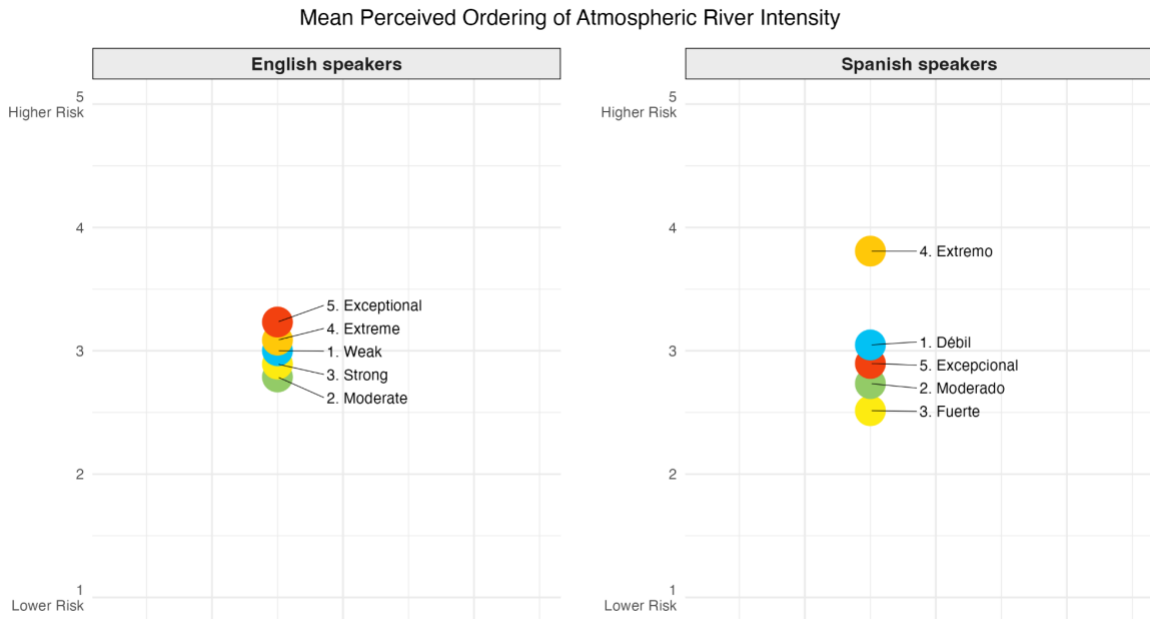
449 To supplement the national findings, we present results from a subsample of participants
 450 residing in the Western NWS region (as defined by NWS n.d.), which experiences the most
 451 frequent and well-documented AR landfalls in the United States. Participants were assigned
 452 to NWS regions based on their self-reported ZIP code. The Western NWS region subsample
 453 includes 260 English speakers from FL25 and 52 Spanish speakers from FLS25. Given the
 454 substantially smaller sample sizes relative to the national sample, these results should be
 455 interpreted with caution and treated as exploratory. Confidence intervals are wider
 456 throughout, and some cells have limited data, particularly for the Spanish subsample.
 457 Nonetheless, these patterns may be of interest to practitioners and researchers focused
 458 specifically on AR-prone regions. We call on researchers to examine this region in deeper
 459 detail in the future so that more generalizable findings can be provided.

460 Fig. A1 presents the weighted distributions of NWS Western Region English and Spanish
 461 speakers' rankings of each AR intensity term across Levels 1–5, mirroring the structure of
 462 Fig. 3 in the main text. Fig. A2 plots the weighted mean perceived rank for each intensity
 463 term among NWS Western Region respondents, following the format of Fig. 4. Fig. A3
 464 presents the weighted distributions of perceived impact (primarily beneficial, equally
 465 beneficial, and hazardous, primarily hazardous, and not sure) for each AR category among
 466 NWS Western Region participants, corresponding to Fig. 5 in the main text. Finally, Fig. A4
 467 presents the AR category thresholds at which NWS Western Region English and Spanish
 468 speakers reported they would begin to consider each of the seven protective actions,
 469 paralleling Fig. 6.

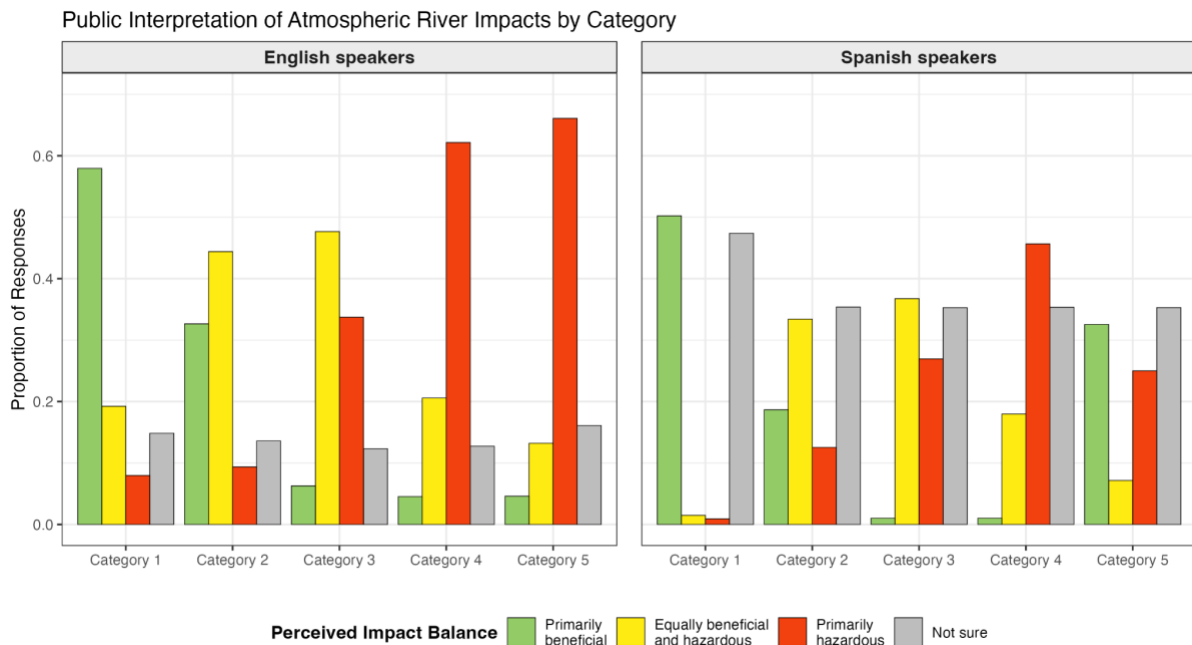


470

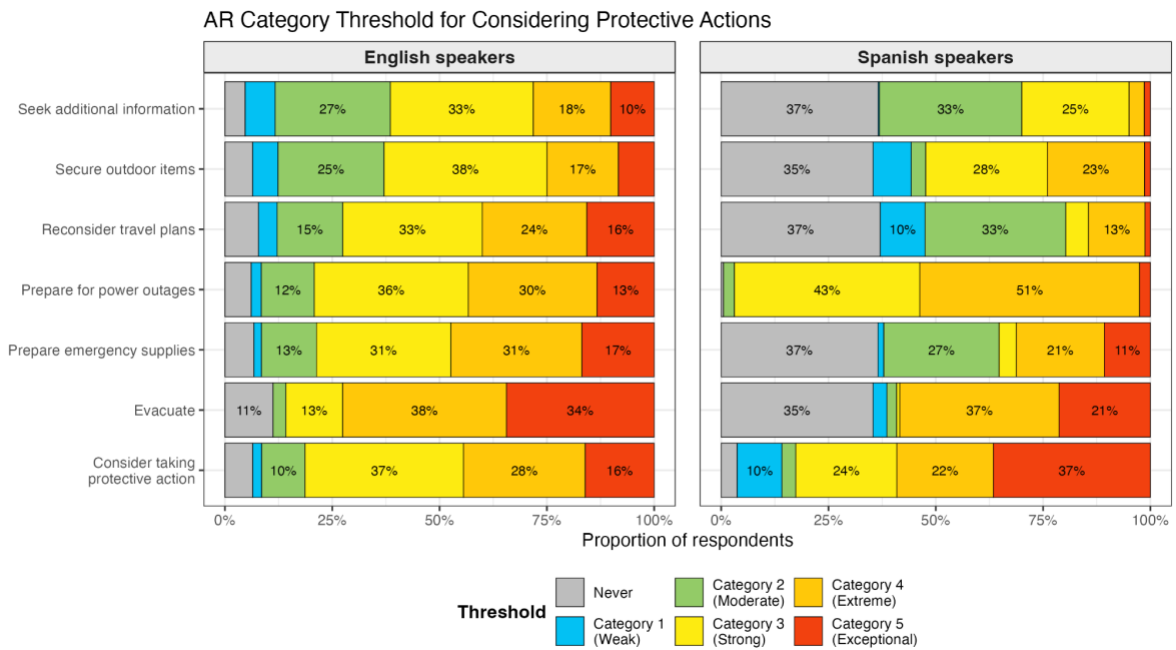
471 **Fig. A1.** Distributions of English and Spanish Western U.S. subsample participants' rankings
 472 of the AR Scale categorical words.



473
 474 **Fig. A2.** Mean placement of each AR Scale intensity term among English and Spanish
 475 Western U.S. subsample participants.



476
 477 **Fig. A3.** Distributions of English and Spanish Western U.S. subsample participants'
 478 characterizations of expected impacts by AR category.



479

480 **Fig. A4.** Stated protective action thresholds by AR category among English and
 481 Western U.S. subsample participants.

482

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