

31 **Abstract**

32 Using behavioral nudges to motivate pro-environmental behavior appeals to program
33 administrators seeking cost-effective ways to increase adoption of environmental practices.
34 However, not all nudges are effective, and reporting when nudges fail is as important as
35 documenting their successes. We used a framed field experiment with 308 adults from the Mid-
36 Atlantic to test the effectiveness of an expert testimonial in encouraging adoption of native plants
37 in residential settings. Though studies have found testimonials to be effective in other contexts,
38 we find that the video testimonial had no effect on residents' willingness to pay for native plants.
39 Our analysis also shows that consumers who are younger, have higher incomes, and use other
40 environmentally friendly practices on their lawns are more likely than other consumers to
41 purchase native plants.

42 **Introduction**

43 Policymakers and program managers are increasingly interested in identifying effective
44 non-monetary behavioral interventions to encourage pro-environmental behavior [1].
45 Collectively referred to as nudges, these interventions can be attractive because they are
46 inexpensive, relatively simple to implement, and preserve individual freedom of choice [2].
47 Growing evidence suggests that nudge-style interventions have the potential to motivate energy
48 and water conservation [3–7], improve recycling efforts [8], and reduce nonpoint source
49 pollution [9].

50 However, gaps remain in our understanding of the effectiveness of specific interventions
51 in motivating pro-environmental behavior [10]. Based on a review of 160 experiment-based
52 studies of environmental interventions, Byerly et al. [10] identified areas in need of future

53 research to inform the design of policies and programs, including testing the effectiveness of
54 interventions used in other contexts in promoting pro-environmental behavior. Behavioral
55 interventions such as information provision [11], peer comparisons [12], social norms [13,14],
56 and framing [15,16] have been extensively studied. Less is known about testimonials, such as
57 opinions and recommendations, provided by experts and influencers who speak positively about
58 a pro-environmental product or practice.

59 Several studies have found that testimonials are effective in some contexts [17–19]. In
60 terms of cost-effectively promoting demand for consumer goods [20–25], they have been called
61 the “workhorse selling tool” that never goes out of vogue [20, p 29]. They have also been
62 identified as successful in promoting positive changes in health-related [17,26,27], child safety
63 [28], and mental health [29] behaviors. Testimonials by experts who are seen as credible have
64 been particularly effective in improving the believability of messages [30,31], and several studies
65 have found that audio-visual testimonials are more effective than text and photos [32,33].

66 A few studies have explored the impacts of testimonial-based interventions on the
67 adoption of pro-environmental behavior. Elgaaied-Gambier et al. [34] and He et al. [35] showed
68 that having an endorser in printed advertisements had a positive impact on the intent of French
69 consumers and students in China, respectively, to choose positive environmental behaviors.
70 Studies that have examined the role of celebrity endorsements have produced mixed results.
71 Olmedo et al. [36] and Ellis et al. [37] found that celebrity endorsements were largely ineffective
72 in inducing pro-environmental behavior, while Ho et al. [38] demonstrated that a celebrity
73 endorsement combined with an information campaign led to a 25% reduction in the use of plastic
74 items among students in Vietnam. These studies provide valuable insights but present challenges
75 when applying their findings to pro-environmental behavior in other countries, such as the

76 United States. We extend the existing body of knowledge by conducting a revealed preference
77 study of U.S. consumers to test the effect of an expert testimonial on willingness to pay (WTP)
78 for native plants, which offer a suite of environmental benefits when planted in residential
79 landscapes.

80 An important issue associated with experiment-based studies of behavioral interventions
81 for pro-environmental behavior is the absence of statistical power analyses to determine adequate
82 sample sizes and adequate randomization [10]. The results of other studies, consequently, can be
83 misleading for policymakers and program managers when they fail to detect existing effects and
84 overstate detected effects [39]. Therefore, randomized controlled experiment designs and
85 sufficient sample sizes are needed to establish causal relationships between specific interventions
86 and pro-environmental behavior [10,40].

87 In this paper, we contribute to the existing literature related to testimonials – broadly
88 defined as opinions and recommendations expressed by experts or influencers encouraging a
89 behavior – by evaluating the effectiveness of an expert testimonial in promoting adoption of pro-
90 environmental behavior using a well-powered randomized experiment design. We test the effect
91 of viewing a video testimonial from an expert on participants’ WTP for native plants in an
92 incentive-compatible framed field experiment by recruiting 308 adults from the U.S. Mid-
93 Atlantic region.

94 Pro-environmental behaviors generally are defined as behaviors that reduce the actors’
95 negative impacts on the environment [41,42]. In this study, we focus on a specific residential
96 pro-environmental behavior – the purchase of native plants. Native plants provide numerous
97 environmental services, including supporting biodiversity [43], and they are an environmentally
98 friendly best management practice for residential landscaping. They require less water and fewer

99 pesticides and fertilizers than do conventional lawns [44], and reduce losses of biodiversity in
100 urban and suburban landscapes by attracting a wide variety of insects and birds [43]. And, like
101 all plants, they sequester carbon [45] and reduce runoff.

102 An important feature of our experiment is that it is non-hypothetical, presenting
103 participants with opportunities to purchase native plants for their own use. This design allows us
104 to examine the effect of an expert's video testimonial in an active market setting in which
105 participants exchange real money for real goods, thus revealing their true preferences [46].

106 **Materials and methods**

107 **Experiment design**

108 We designed an incentive-compatible single-bounded dichotomous-choice framed field
109 experiment to test the effect of a video testimonial on WTP for Husker Red (*Penstemon*
110 *digitalis*), a perennial plant native to the U.S. Mid-Atlantic region that was recommended by a
111 native plant expert at the local land grant institution. The experiment, which took approximately
112 15 minutes to complete, was conducted in April and May 2018 at three locations in the U.S.
113 Mid-Atlantic region – a large community event, a super-regional shopping mall frequented by
114 more than 20 million shoppers each year, and an ice cream shop. The locations were selected
115 because they provided access to a diverse pool of participants.

116 Signs and flyers were used to recruit potential participants 22 years of age or older who
117 make management decisions about the landscape around their home in a two-part experiment.
118 Prior to making those decisions, each participant was endowed with a \$10 participation payment
119 that was theirs to keep – they could also use this money to purchase plants. Participants were
120 then assigned an iPad Pro with headphones after reading and signing a consent statement to

121 participate in the experiment. Participants completed the experiment asynchronously and were
122 not permitted to communicate with other participants present at the same time. The experiment
123 (control and treatment versions), the choice questions, and a short survey were presented
124 electronically via the iPad screens.

125 To test the effect of the expert testimonial on their WTP for plants, participants were
126 randomly assigned to either the non-video control group or the video-testimonial treatment
127 group. Before making their WTP choices, they received written information about the benefits of
128 native plants (see Fig 1). Participants in the treatment group viewed the 90-second video
129 testimonial in which the benefits of native plants were described by Dr. Doug Tallamy, a well-
130 known professor of entomology and wildlife ecology at a local land-grant university, who is a
131 highly regarded expert in the field and author of a best-selling book about conservation and
132 landscape ecology. A transcript of the video testimonial and a web link to the video can be found
133 in S1 Appendix A of this paper. In the experiment, participants in the treatment group were
134 required to watch the entire video before they could proceed with the experiment.

135 **Fig 1. Written information provided to participants about the benefits of native plants.**

136 During the first portion of the experiment, participants were given the opportunity to
137 purchase a bundle of three ready-to-plant landscape plugs of Husker Red. The iPads presented
138 participants with a description of the plant and a picture of how it would look once fully grown
139 (see Fig 2). Participants were asked to indicate their willingness to purchase the bundle at a
140 stated price in three dichotomous-choice (Yes/No) questions that presented different prices: \$0,
141 \$3, and \$6 per bundle. We included a free plant bundle (\$0 price) decision because the second
142 stage of the experiment, which is not reported here, was designed to analyze how self-reported

143 plantings were affected by receiving the plants for free versus purchasing them. The prices were
144 presented in random order to avoid potential ordering effects.

145 **Fig 2. Description of the native plant offered to experiment participants.**

146 To ensure incentive-compatibility, participants were told that only one of the three
147 purchase decisions would be binding, and that the binding decision would be randomly selected
148 at the end of the study to determine participants' take-home cash and whether they purchased the
149 native plants. This mechanism ensured that participants' dominant strategy was to answer 'Yes'
150 whenever the posted price did not exceed their maximum WTP for the bundle. After answering
151 the WTP questions, participants completed a short survey that collected information on their
152 socio-demographic characteristics and their lawn care practices.

153 Once they completed the survey, the program randomly selected one of the prices
154 (decisions) as binding and notified participants of their outcomes. Participants who indicated
155 'Yes' in the selected decision purchased the bundle of plants at that price, which was deducted
156 from the \$10 participation payment; they received the plant bundle and the balance of the
157 participation payment. Participants who indicated 'No' did not purchase plants and received the
158 entire \$10 participation payment. As participants exited the experiment area, they returned the
159 iPads and received their take-home funds and plants purchased, if any. A detailed experiment
160 protocol and copy of the survey are provided in S2 Appendix B.

161 Before conducting the experiment, we performed an ex-ante power analysis to determine
162 the required sample size for the full experiment, which consisted of three treatment arms and two
163 stages (in this paper, we present results of one treatment from the first stage of the experiment).
164 The power analysis was conducted via simulation following the approach described by Feiveson

165 [47] so that we could model the multi-stage structure of the experiment and account for some
166 participants exiting the experiment after each stage. To implement the simulation, we used
167 valuations of native plants collected via an experimental auction by members of our research
168 team [15]. The final experimental design required 345 participants to reliably detect the effect of
169 the video-testimonial on WTP for native plants, assuming a standardized effect size of 0.30 with
170 80% power (i.e., $\beta = 0.80$). Because we obtained data from only 308 participants, we also
171 conducted an ex-post power analysis to understand the implications for our analysis. We present
172 additional details about the ex-post power analysis in the results section.

173 **Data**

174 In the experiment, we collected valid data from 309 participants. One participant was dropped
175 because of invalid data entry, resulting in a final sample of 308 adult individuals. According to
176 participants' self-reported zip codes, about 72.8% of the participants resided in Delaware, 9.1%
177 of the participants resided in Pennsylvania, and 7.4% of the participants resided in Maryland.
178 Table 1 provides summary statistics of the participants' socio-demographic characteristics. On
179 average, participants in the experiment were 41 years old, 44% were male, and 42% had one or
180 more children under 18 years old in their households. Approximately 66% of the participants
181 were employed, and about one-third of the participant households fell into each annual income
182 category: less than \$50,000; \$50,000 to \$100,000; and greater than \$100,000. In terms of
183 education, 67% of the participants had undergraduate and/or graduate degrees. Most lived in
184 urban areas (87%) and owned their homes (68%). About 84% of the participants indicated that
185 they used environmentally friendly practices to maintain their lawns.

186 Two sample t-tests showed that the control and treatment groups were balanced in terms
187 of most of the socio-demographic characteristics. However, compared to the control group, the

188 treatment group included more males, participants with children, and annual incomes between
 189 \$50,000 and \$100,000.

190 A comparison of the sample to the general population of Delaware, where most of the
 191 participants resided (72.8%), showed that our sample was broadly representative of the state's
 192 population in terms of age, gender, and income. Our sample exceeded estimates for Delaware's
 193 population in terms of employment rate, number of children in the household, and education level.

194 **Table 1. Summary statistics for socio-demographic variables by treatment.**

Variable	Definition	Entire Sample	Control	Treatment	Delaware (2018)^a
Age	Participant's age in years in 2018	40.70 (15.00)	41.12 (15.35)	40.28 (14.65)	40.20 (Median)
Male	Equals 1 for male participants	0.438 (0.496)	0.494 (0.500)	0.383 (0.487)	0.484
Children	Equals 1 when a child under 18 resided in the household	0.416 (0.493)	0.370 (0.483)	0.461 (0.499)	0.252
Employed	Equals 1 when participant was employed	0.659 (0.474)	0.662 (0.473)	0.656 (0.476)	0.586
Income _{<\$50K}	Equals 1 when participant's income was less than \$50,000	0.315 (0.465)	0.344 (0.476)	0.286 (0.452)	0.383
Income _{\$50K-\$100K}	Equals 1 when participant's income was between \$50,000 and \$100,000	0.354 (0.478)	0.299 (0.458)	0.409 (0.492)	0.322
Income _{>\$100K}	Equals 1 when participant's income was greater than \$100,000	0.331 (0.471)	0.357 (0.480)	0.305 (0.461)	0.296
Education level	Equals 1 for participants who had a bachelor, graduate, or professional degree	0.672 (0.470)	0.682 (0.466)	0.662 (0.473)	0.314
Urban	Equals 1 for participants living in urban areas	0.873 (0.333)	0.890 (0.314)	0.857 (0.350)	
Homeowner	Equals 1 for participants who owned their homes	0.679 (0.467)	0.662 (0.473)	0.695 (0.461)	
Lawnsize _{small}	Equals 1 when lawn size was less than 0.25 acres or participants did not have a lawn	0.425 (0.495)	0.422 (0.494)	0.429 (0.495)	
Lawnsize _{medium}	Equals 1 when lawn size was between 0.25 and 1.0 acres	0.448 (0.498)	0.435 (0.496)	0.461 (0.499)	

Lawnsizelarge	Equals 1 when lawn size was greater than 1.0 acres	0.127 (0.333)	0.143 (0.350)	0.110 (0.314)
Lawn practice	Equals 1 when participants use environmentally friendly practices at home for the lawn	0.844 (0.365)	0.831 (0.375)	0.857 (0.350)
N		308	154	154 949,495

Mean values are in bold. Standard deviations are in parentheses.

Although t-tests showed that only male, children, and income_{\$50K-\$100K} are not balanced between the control and treatment groups, joint orthogonality tests showed that none of the variables are jointly orthogonal to the treatment.

^a Source: 2018 American Community Survey: www.census.gov/acs/www/data/data-tables-and-tools/data-profiles/2018/.

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196 Analysis

197 The 308 participants in the experiment each made three ‘Yes/No’ purchase decisions, resulting in
 198 924 observations. Given the binary nature of the decisions, we used a logistic regression to
 199 analyze the impact of the testimonial on purchases. Since any effect of the testimonial was
 200 constant for the three decisions, we used a random effects logistic model with the following
 201 specification [48]:

$$P(D = 1) = \frac{\exp(\beta_0 + \beta_1 P + \beta_2 \textit{Testimonial} + \boldsymbol{\gamma}'\mathbf{Z} + \varepsilon)}{1 + \exp(\beta_0 + \beta_1 P + \beta_2 \textit{Testimonial} + \boldsymbol{\gamma}'\mathbf{Z} + \varepsilon)} \quad (1)$$

202 where D is a dummy variable indicating purchase decisions, $D = 1$ represents a ‘Yes’ decision,
 203 and P is the posted price. *Testimonial* is a treatment dummy variable indicating whether a
 204 participant watched the video testimonial. Therefore, β_2 is the coefficient of interest that shows
 205 the treatment effect. \mathbf{Z} is a vector of the socio-demographic variables presented in Table 1, and ε
 206 is the individual stochastic error term. We estimated two models: a simple model that did not
 207 include the socio-demographic variables and a more complex one that included all of the socio-

208 demographic variables to control for unbalanced variables between the control and treatment
209 groups.

210 Using coefficient estimates from the logistic regression, we calculated the mean WTP for
211 native plants following [49]:

$$WTP = \frac{1}{\hat{\beta}_1} (\hat{\beta}_0 + \hat{\beta}_2 Testimonial + \hat{\gamma}' \bar{\mathbf{Z}}) \quad (2)$$

212 in which $Testimonial = 1$ for WTP by treatment participants and $Testimonial = 0$ for WTP by
213 control group participants. $\bar{\mathbf{Z}}$ includes the mean values (e.g., age, income) of the socio-
214 demographic variables for the control and treatment groups. The models were estimated using
215 maximum likelihood, and mean values and standard errors of WTP were calculated using the
216 delta method in Stata.

217 **Results**

218 We first examine participants' purchasing decisions by looking at differences in the percentage
219 of participants who were willing to purchase native plants at each posted price (percentage of
220 'Yes' responses). As shown in Fig 3, the percentage of participants willing to purchase the native
221 plants decreased as the posted price increased. Interestingly, only about 80% were willing to take
222 the free native plants (price of \$0). The fact that some participants did not want the plants likely
223 reflected both their landscaping preferences and being unwilling to assume the cost and/or care
224 of the plants. About 43.5% and 27.3% of participants were willing to purchase the plants at \$3
225 and \$6, respectively.

226 **Fig 3. Percentage of 'yes' responses at each price.**

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228 As Fig 3 shows, the percentage of participants willing to take the free plants in the
 229 control and treatment groups is identical (79.2%). At a price of \$3, more treatment participants
 230 (51.9%) than control participants (43.5%) were willing to purchase the native plants. At \$6,
 231 slightly more control participants (27.3%) than treatment participants (25.3%) were willing to
 232 purchase the plants.

233 Table 2 reports the coefficients estimated from the random effects logistic regression
 234 used to identify factors that influenced participants' likelihood of purchasing native plants. We
 235 present the results for both the simple model that estimated only the treatment effect (Model 1)
 236 and the expanded model that included the full set of socio-demographic characteristics
 237 (Model 2). Based on the log pseudolikelihood, Model 2 slightly outperformed Model 1 so we
 238 draw our conclusions from the results from Model 2.

239 **Table 2. Random effects logistic regression results.**

	Model 1		Model 2	
	Coeff.	S.E.	Coeff.	S.E.
Price	-0.655***	(0.067)	-0.656***	(0.067)
Testimonial	0.152	(0.305)	0.112	(0.307)
Age			-0.038***	(0.011)
Male			0.159	(0.305)
Children			-0.299	(0.314)
Employed			-0.232	(0.318)
Income\$50K-\$100K			0.932**	(0.402)
Income>\$100K			0.818*	(0.448)
Urban			-0.219	(0.548)
Education level			0.304	(0.332)
Homeowner			0.524	(0.362)
Lawnsize _{medium}			-0.099	(0.342)
Lawnsize _{large}			0.792	(0.504)

Lawn practice			0.761*	(0.402)
Constant	1.987***	(0.302)	2.076**	(0.807)
Log pseudolikelihood	-504.192		-491.869	
N	927		927	

***, **, and * denote statistical significance at the 1%, 5%, and 10% level respectively.

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The results of the regression indicate that price had a negative effect on opting to purchase the plants. This result is expected and is evident in Fig 3. We also find that the coefficient of the variable of interest, *Testimonial*, is positive but not significantly different from zero. Thus, we find that the video testimonial had no significant effect on participants' decisions about purchasing native plants.

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In terms of socio-demographic characteristics, we find that older participants were less likely to purchase native plants than younger participants and that higher income (\$50,000 or more per year) participants were more likely to purchase native plants than were lower income (less than \$50,000 per year) participants. We further find that participants who already used environmentally friendly practices to care for their lawns were more likely to purchase native plants. This result is intuitive since planting native species represents pro-environmental behavior.

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Using equation 2 and the coefficient estimates from Model 2, we calculated the mean WTP for native plants for the control and treatment groups and report the results in Fig 4. The mean WTP for native plants by treatment participants was \$0.26 higher than the mean WTP by control participants, which (consistent with the results of our earlier analysis) is not statistically significant. Thus, these results further indicate that the video testimonial did not have a significant impact on participants' WTP for native plants.

259 **Fig 4. Mean willingness to pay for native plants.** Note: Dots represent mean WTP estimates.
260 Vertical lines represent 95% confidence intervals calculated using the delta method.

261 Collectively, our findings from the random effects logistic model and analysis of mean
262 WTP values show that the behavioral intervention of providing participants with a video
263 testimonial by an expert did not result in significant changes in participants' preferences for the
264 pro-environmental behavior of buying native plants.

265 Failing to reject the null hypothesis of no treatment effect implies that either there is no
266 treatment effect or the experiment did not have enough statistical power to detect the effect [50].
267 We conducted an ex-post power analysis to determine the size of the testimonial effect that we
268 could detect with 80% power using our sample of 308 participants. The results of the power
269 analysis vary depending on the underlying assumptions made about the distribution of WTP
270 values. Our assumptions for the power analysis were guided by two previous studies of WTP for
271 native plants. Using an experimental auction, Yue et al. [51] estimated the mean and standard
272 deviation for plants labeled native/noninvasive as \$2.79 and \$2.90, respectively, with 5% of the
273 bids at \$0. Li et al. [15] estimated the mean and standard deviation to be \$4.55 and \$4.98,
274 respectively, with 20% of the bids at \$0. The results of these two studies suggest that the mean
275 and standard deviation of values are similar in magnitude and that the distribution of values is
276 skewed to the left. Considering the value distributions from these two studies, we conduct two
277 ex-post power analyses. First, we calculate the minimum detectable effect size using Stata's
278 'power' command. Assuming a mean and standard deviation of \$3.00, the minimum detectable
279 effect size is \$0.96, which equates to a standardized effect size of about 0.32. Since this
280 calculation does not account for the structure of our data and the fact that we are not eliciting
281 WTP values directly, we also use a simulation approach to calculate our power using the

282 experimental auction bids elicited in Li et al. [15]. From this analysis, we find that our study is
283 designed to detect an effect of \$1.50 with 80% power.

284 Based on our power analyses, we cannot rule out the possibility that the testimonial has
285 an effect below the \$0.96 to \$1.50 range. A larger sample size would allow us to detect smaller
286 effects at the cost of recruiting and paying more participants. Researchers must weigh these costs
287 against the potential benefits of detecting smaller effects. For example, our ex-post power
288 analysis using the same simulation showed that we would have needed a sample of 6,000
289 participants to detect a \$0.26 difference in mean WTP. Thus, it is possible that the video
290 testimonial induced differences in mean WTP that were smaller than our study was able to
291 detect, but such differences would not necessarily be economically meaningful for policy and
292 program designs aimed at promoting pro-environmental behavior.

293 **Discussion**

294 As policymakers, program managers, and practitioners seek new ways of integrating nudge-style
295 behavioral interventions into policies and programs designed to promote pro-environmental
296 behavior, a more robust literature is needed that not only identifies what does work, but
297 importantly also identifies things that do not work or have limited impacts. This study examines
298 the effectiveness of a testimonial (broadly defined as opinions and recommendations expressed
299 by experts or influencers encouraging a behavior) by an environmental expert. Despite effective
300 and widespread use of testimonials to market consumer goods and influence health-related
301 decisions [17,20–27,29], little research has been conducted on using testimonials to encourage
302 pro-environmental choices. We test this behavioral intervention in an incentive-compatible
303 experiment involving actual consumer decisions about purchasing a native plant for their
304 residential lawns as an environmentally beneficial landscaping choice. We compare the WTP for

305 native plants between participants who viewed a video testimonial from a renowned expert in the
306 field about the benefits of native plants and participants who did not view the video testimonial.

307 Contrary to studies of testimonials used in other contexts [17,26,32], we find that the
308 expert's video testimonial had no significant identifiable effect on WTP for native plants.
309 Consequently, in pro-environmental contexts, testimonial-based interventions alone are not
310 necessarily sufficient to motivate changes in behavior.

311 We further find that consumers who are younger, have relatively higher incomes, and
312 already use other environmentally friendly practices for their lawns are more likely than other
313 consumers to purchase native plants. These findings are consistent with previous studies
314 exploring the determinants of pro-environmental behavior (see Blankenberg and Alhusen [52]
315 for a review).

316 Our results make several important contributions. First, we extend the literature
317 examining non-monetary nudges on adoption of pro-environmental behavior by analyzing the
318 effect of expert testimonials using a carefully designed, revealed preference experiment.
319 Publishing null results of sufficiently powered economic studies is critical to efforts to address
320 publication bias that occurs when only statistically significant findings are reported [53]. Finally,
321 our findings have practical implications, particularly for environmental organizations considering
322 using expert testimonials to increase engagement in their programs. Our results show that such
323 organizations should consider whether the cost of creating and disseminating testimonials is
324 worth the expected benefit, which could be quite small based on the results of our study.

325 While our results make valuable contributions to the literature, we must acknowledge
326 several limitations. We test the influence of a testimonial in a specific context – purchase of

327 native plants. Future research could determine whether testimonials can influence other types of
328 pro-environmental behavior. Furthermore, future studies could test the effect of combining
329 testimonials with other non-monetary behavioral interventions and/or financial incentives.
330 Finally, studies conducted with larger samples could potentially detect smaller effects of
331 testimonials alone.

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339 **References**

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341 1. Sunstein CR. Green defaults can combat climate change. *Nature Human Behaviour* [Internet].
342 2021;5(5):548–549. Available from: <https://doi.org/10.1038/s41562-021-01071-2>

343

344 2. Thaler RH, Sunstein CR. *Nudge: Improving decisions about health, wealth, and happiness*.
345 New Haven: Yale University Press; 2008.

346

347 3. Allcott H, Kessler JB. The welfare effects of nudges: A case study of energy use social
348 comparisons. *American Economic Journal: Applied Economics* [Internet]. 2019;11(1):236–276.
349 Available from: <https://www.aeaweb.org/articles?id=10.1257/app.20170328>

350

351 4. Asensio OI, Delmas MA. Nonprice incentives and energy conservation. *Proceedings of the*
352 *National Academy of Science* [Internet]. 2015 Feb 10;112(6):E510 LP-E515. Available from:
353 <http://www.pnas.org/content/112/6/E510.abstract>

354

355 5. Delmas MA, Fischlein M, Asensio OI. Information strategies and energy conservation
356 behavior: A meta-analysis of experimental studies from 1975 to 2012. *Energy Policy* [Internet].
357 2013;61:729–739. Available from:

358 <https://www.sciencedirect.com/science/article/pii/S0301421513004643>

359

360 6. Brent DA, Cook JH, Olsen S. Social comparisons, household water use, and participation in
361 utility conservation programs: Evidence from three randomized trials. *Journal of the Association*
362 *of Environmental Resource Economists* [Internet]. 2015 Nov 18;2(4):597–627. Available from:
363 <https://doi.org/10.1086/683427>

364

365 7. Ferraro PJ, Price MK. Using nonpecuniary strategies to influence behavior: Evidence from a
366 large-scale field experiment. *Review of Economics and Statistics*. 2013;95(1):64–73.

367

368 8. Kirakozian A. One without the other? Behavioural and incentive policies for household waste
369 management. *Journal of Economic Surveys* [Internet]. 2016 Jul 1;30(3):526–551. Available
370 from: <https://doi.org/10.1111/joes.12159>

371

372 9. Fleming PM, Palm-Forster LH, Kelley LE. The effect of legacy pollution information on
373 landowner investments in water quality: Lessons from economic experiments in the field and the
374 lab. *Environmental Resource Letters* [Internet]. 2021;16(4):45006. Available from:
375 <http://dx.doi.org/10.1088/1748-9326/abea33>

- 376
377 10. Byerly H, Balmford A, Ferraro PJ, Hammond Wagner C, Palchak E, Polasky S, et al.
378 Nudging pro-environmental behavior: Evidence and opportunities. *Frontiers in Ecology and the*
379 *Environment* [Internet]. 2018 Apr 1;16(3):159–168. Available from:
380 <https://doi.org/10.1002/fee.1777>
381
- 382 11. Abrahamse W. Behaviour change interventions. In: Abrahamse W, editor. *Encouraging Pro-*
383 *environemntal behaviour*. London: Academic Press; 2019. pp. 27–48.
384
- 385 12. Wu S, Palm-Forster LH, Messer KD. Impact of peer comparisons and firm heterogeneity on
386 nonpoint source water pollution: An experimental study. *Resource Energy Economics* [Internet].
387 2021;63:101142. Available from:
388 <https://www.sciencedirect.com/science/article/pii/S092876551830318X>
389
- 390 13. Farrow K, Grolleau G, Ibanez L. Social norms and pro-environmental behavior: A review of
391 the evidence. *Ecological Economics* [Internet]. 2017; 140:1–13. Available from:
392 <https://www.sciencedirect.com/science/article/pii/S0921800915301543>
393
- 394 14. Schubert C. Green nudges: Do they work? Are they ethical? *Ecological Economics*
395 [Internet]. 2017;132:329–342. Available from:
396 <https://www.sciencedirect.com/science/article/pii/S0921800916301860>
397
- 398 15. Li T, Fooks JR, Messer KD, Ferraro PJ. A field experiment to estimate the effects of
399 anchoring and framing on residents’ willingness to purchase water runoff management
400 technologies. *Resource Energy Economics* [Internet]. 2021; 63:101-107. Available from:
401 <https://www.sciencedirect.com/science/article/pii/S0928765518303129>
402
- 403 16. Savchenko OM, Li T, Kecinski M, Messer KD. Does food processing mitigate consumers’
404 concerns about crops grown with recycled water? *Food Policy* [Internet]. 2019;88:101748.
405 Available from: <https://www.sciencedirect.com/science/article/pii/S0306919219305652>
406
- 407 17. Durantini MR, Albarracín D, Mitchell AL, Earl AN, Gillette JC. Conceptualizing the
408 influence of social agents of behavior change: A meta-analysis of the effectiveness of HIV-
409 prevention interventionists for different groups. *Psychological Bulletin*. 2006; 132:212–248.
410
- 411 18. Durkin S, Brennan E, Wakefield M. Mass media campaigns to promote smoking cessation
412 among adults: An integrative review. *Tobacco Control* [Internet]. 2012 Aug 26;21(2):127–138.
413 Available from: <http://www.jstor.org/stable/41516004>
414
- 415 19. Igartua J-J, Rodríguez-Contreras L. Narrative voice matters! Improving smoking prevention
416 with testimonial messages through identification and cognitive processes. *International Journal*
417 *of Environmental Research and Public Health*. 2020; 17:72811. Available from:
418 <http://doi.org.10.3390/ijerph17197281>
419

420 20. Martin BAS, Wentzel D, Tomczak T. Effects of susceptibility to normative influence and
421 type of testimonial on attitudes toward Print advertising. *Journal of Advertising* [Internet]. 2008
422 Apr 1; 37(1):29–43. Available from: <https://doi.org/10.2753/JOA0091-3367370103>
423

424 21. Shimp TA, Wood SL, Smarandescu L. Consumer testimonials as self-generated
425 advertisements: Evaluative reconstruction following product usage. *MSI Reports*. 2005;2(05–
426 109):91–113.
427

428 22. Bly RW. More tips on using testimonials [Internet]. 2005 [cited 2019 Apr 17]. Available
429 from: <http://marketingtoday.com/marcom/testi2.htm>
430

431 23. Wright AA, Lynch JG. Communication effects of advertising versus direct experience when
432 both search and experience attributes are present. *Journal of Consumer Research* [Internet]. 1995
433 Aug 26;21(4):708–718. Available from: <http://www.jstor.org/stable/2489726>
434

435 24. Shapiro S, Spence MT. Factors affecting encoding, retrieval, and alignment of sensory
436 attributes in a memory-based brand choice task. *Journal of Consumer Research* [Internet]. 2002
437 Mar 1;28(4):603–617. Available from: <https://doi.org/10.1086/338204>
438

439 25. Kemp E, Williams KH, Porter M. Hope across the seas: The role of emotions and risk
440 propensity in medical tourism advertising. *International Journal of Advertising* [Internet]. 2015
441 Aug 8;34(4):621–40. Available from: <https://doi.org/10.1080/02650487.2015.1024385>
442

443 26. Rollins B, Huh J, Bhutada N, Perri M. Effects of endorser type and testimonials in direct-to-
444 consumer prescription drug advertising (DTCA). *International Journal of Pharmaceutical and*
445 *Healthcare Marketing* [Internet]. 2021 Jan 1;15(1):1–17. Available from:
446 <https://doi.org/10.1108/IJPHM-06-2019-0042>
447

448 27. Dillard AJ, Main JL. Using a health message with a testimonial to motivate colon cancer
449 screening: Associations with perceived identification and vividness. *Health Education &*
450 *Behavior* [Internet]. 2013 Jan 25;40(6):673–82. Available from:
451 <https://doi.org/10.1177/1090198112473111>
452

453 28. Shen J, Pang S, Schwebel DC. Evaluation of a drowning prevention program based on
454 testimonial videos: A randomized controlled trial. *Journal of Pediatric Psychology* [Internet].
455 2016 Jun 1;41(5):555–565. Available from: <https://doi.org/10.1093/jpepsy/jsv104>
456

457 29. Apolinário-Hagen J, Harrer M, Dederichs M, Fritsche L, Wopperer J, Wals F, et al.
458 Exploring the influence of testimonial source on attitudes towards e-mental health interventions
459 among university students: Four-group randomized controlled trial. *PLoS One* [Internet]. 2021
460 May 26;16(5):e0252012. Available from: <https://doi.org/10.1371/journal.pone.0252012>
461

462 30. Tedeschi JT. *The social influence processes*. Chicago: Aldine·Atherton; 1972.
463

- 464 31. Maddux JE, Rogers RW. Effects of source expertness, physical attractiveness, and
465 supporting arguments on persuasion: A case of brains over beauty. *Journal of Personality and*
466 *Social Psychology*. 1980; 39:235–244.
467
- 468 32. Braverman J. Testimonials versus informational persuasive messages: The moderating effect
469 of delivery mode and personal involvement. *Communication Research* [Internet]. 2008 Aug
470 4;35(5):666–694. Available from: <https://doi.org/10.1177/0093650208321785>
471
- 472 33. Appiah O. Rich media, poor media: The impact of audio/video vs. text/picture testimonial
473 ads on browsers' evaluations of commercial web sites and online products. *Journal of Current*
474 *Issues & Research in Advertising* [Internet]. 2006 Mar 1;28(1):73–86. Available from:
475 <https://doi.org/10.1080/10641734.2006.10505192>
476
- 477 34. Elgaaied-Gambier L, Monnot E, Reniou F. Using descriptive norm appeals effectively to
478 promote green behavior. *Journal of Business Research* [Internet]. 2018;82:179–191. Available
479 from: <https://www.sciencedirect.com/science/article/pii/S014829631730351X>
- 480 35. He H, Fu J, Li X, Guo R. The interplay between endorser social status and normative appeals
481 on the endorsement effectiveness of pro-environmental behaviors. *PLoS One* [Internet]. 2019 Jan
482 15;14(1):e0210699. Available from: <https://doi.org/10.1371/journal.pone.0210699>
483
- 484 36. Olmedo A, Milner-Gulland EJ, Challender DWS, Cugnière L, Dao HTT, Nguyen LB, et al.
485 A scoping review of celebrity endorsement in environmental campaigns and evidence for its
486 effectiveness. *Conservation Science & Practice* [Internet]. 2020 Oct 1;2(10):e261. Available
487 from: <https://doi.org/10.1111/csp2.261>
488
- 489 37. Ellis S, Savchenko O, Messer KD. Mitigating stigma associated with recycled water.
490 *American Journal of Agricultural Economics*. 2021 (forthcoming). Available from:
491 <https://doi.org/10.1111/ajae.12256>
492
- 493 38. Ho T, Nie Z, Alpizar F, Carlsson F, Pham KN. Celebrity endorsement in promoting pro-
494 environmental behavior. *EfD Discussion Paper 20-25*. Gothenburg, Sweden: Environment for
495 Development; 2020.
496
- 497 39. Rosch S, Raszap Skorbiansky S, Weigel C, Messer KD, Hellerstein D. Barriers to using
498 economic experiments in evidence-based agricultural policymaking. *Applied Economic*
499 *Perspectives and Policy* [Internet]. 2021 Jun 1;43(2):531–555. Available from:
500 <https://doi.org/10.1002/aep.13091>
501
- 502 40. Palm-Forster LH, Ferraro PJ, Janusch N, Vossler CA, Messer KD. Behavioral and
503 experimental agri-environmental research: Methodological challenges, literature gaps, and
504 recommendations. *Environmental Resource Economics* [Internet]. 2019;73(3):719–42. Available
505 from: <https://doi.org/10.1007/s10640-019-00342-x>
506
- 507 41. Stern PC. New environmental theories: Toward a coherent theory of environmentally
508 significant behavior. *Journal of Social Issues* [Internet]. 2000 Jan 1;56(3):407–424. Available
509 from: <https://doi.org/10.1111/0022-4537.00175>

- 510
511 42. Kollmuss A, Agyeman J. Mind the Gap: Why do people act environmentally and what are
512 the barriers to pro-environmental behavior? Environmental Education Research [Internet]. 2002
513 Aug 1;8(3):239–260. Available from: <https://doi.org/10.1080/13504620220145401>
514
- 515 43. Burghardt KT, Tallamy DW, Gregory Shriver W. Impact of native plants on bird and
516 butterfly biodiversity in suburban landscapes. Conservation Biology [Internet]. 2009 Feb
517 1;23(1):219–24. Available from: <https://doi.org/10.1111/j.1523-1739.2008.01076.x>
518
- 519 44. EPA. Greenacres: Landscaping with native plants [Internet]. 2003. Available from:
520 <https://archive.epa.gov/greenacres/web/html/index.html>
- 521 45. EPA. Landscaping with native plants [Internet]. 2002. Available from:
522 [https://www.waterfordmi.gov/DocumentCenter/View/587/Landscaping-with-Native-Plants-](https://www.waterfordmi.gov/DocumentCenter/View/587/Landscaping-with-Native-Plants-PDF?bidId=)
523 [PDF?bidId=](https://www.waterfordmi.gov/DocumentCenter/View/587/Landscaping-with-Native-Plants-PDF?bidId=)
524
- 525 46. Lusk JL, Shogren JF. Experimental auctions: Methods and applications in economic and
526 marketing research—Quantitative Methods for Applied Economics and Business Research.
527 Cambridge: Cambridge University Press; 2007. Available from:
528 [https://www.cambridge.org/core/books/experimental-](https://www.cambridge.org/core/books/experimental-auctions/6B2CD58A523F3B4591783F0947627347)
529 [auctions/6B2CD58A523F3B4591783F0947627347](https://www.cambridge.org/core/books/experimental-auctions/6B2CD58A523F3B4591783F0947627347)
530
- 531 47. Feiveson AH. Power by simulation. Stata Journal [Internet]. 2002 Jun 1;2(2):107–24.
532 Available from: <https://doi.org/10.1177/1536867X0200200201>
533
- 534 48. Wooldridge JM. Introductory econometrics: A modern approach. 7th edition. Boston:
535 Cengage Learning; 2019.
536
- 537 49. Hanemann WM. Welfare evaluations in contingent valuation experiments with discrete
538 responses. American Journal of Agricultural Economics [Internet]. 1984 Aug 1;66(3):332–341.
539 Available from: <https://doi.org/10.2307/1240800>
540
- 541 50. Glennerster R, Takavarasha K. Running randomized evaluations: A practical guide
542 [Internet]. Princeton: Princeton University Press; 2013. Available from:
543 <http://www.jstor.org.udel.idm.oclc.org/stable/j.ctt4cgd52.10>
544
- 545 51. Yue C, Hurley TM, Anderson N. Do native and invasive labels affect consumer willingness
546 to pay for plants? Evidence from experimental auctions. Agricultural Economics [Internet]. 2011
547 Mar 1;42(2):195–205. Available from: <https://doi.org/10.1111/j.1574-0862.2010.00510.x>
548
- 549 52. Blankenberg AK, Alhusen H. On the determinants of pro-environmental behavior: A
550 literature review and guide for the empirical economist. 2019. Cege Discussion Papers No. 350,
551 University of Goettingen, Goettingen, Germany.
552
- 553 53. Ioannidis JPA, Stanley TD, Doucouliagos H. The power of bias in economics research.
554 Economics Journal [Internet]. 2017 Oct 1;127(605):F236–65. Available from:
555 <https://doi.org/10.1111/eoj.12>

556 **Supporting Information**

557

558 **S1 Appendix A. Transcript of the video testimonial provided by an expert.**

559 **S2 Appendix B. Experiment instructions and survey questions.**