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- US-China differences in cognition and perception across 12 tasks: Replicability, robustness,
- and within-culture variation
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Abstract

Cultural differences between the US and China have been investigated using a broad array 17 of psychological tasks measuring differences between cognition, language, perception, and 18 reasoning. We examine the robustness of several classic experimental paradigms in 19 cross-cultural psychology. Using online convenience samples of adults, we conducted two 20 large-scale replications of 12 tasks previously reported to show cross-cultural differences. 21 Our results showed a heterogeneous pattern of successes and failures: five tasks yielded 22 robust cultural differences across both experiments, while six showed no difference between 23 cultures, and one showed a small difference in the opposite direction. We observed moderate reliability in all of the multi-trial tasks, but there was little shared variation between tasks. Additionally, we did not see within-culture variation across a range of demographic factors in our samples. Finally, as in prior work, cross-cultural differences in cognition (in those tasks showing differences) were not strongly related to explicit measures 28 of cultural identity and behavior. All of our tasks, data, and analyses are available openly 29 online (and reusable by future researchers), providing a foundation for future studies that 30 seek to establish a robust and replicable science of cross-cultural difference. 31

Keywords: replication; cross-cultural differences

Word count: FIXME

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 and within-culture variation

36 Introduction

Cross-cultural differences are a striking part of the broader landscape of human 37 variation. Differences in values and behavior across cultures are obvious to even a casual 38 observer, and researchers have attempted to quantify these differences via a wide range of 39 measures. Comparisons between Western and East Asian cultures have been especially well-researched, with differences attested in a wide range of cognitive domains, including visual attention (Chua, Boland, & Nisbett, 2005; Ji, Peng, & Nisbett, 2000; Waxman et al., 2016), executive function (Sabbagh, Xu, Carlson, Moses, & Lee, 2006; Tan, 2020), language learning (Chan et al., 2011, 2011; Tardif, 1996; Waxman et al., 2016), relational reasoning (Carstensen et al., 2019; Cheng, 2020; Richland, Chan, Morrison, & Au, 2010; Su, 2020), similarity judgments (Ji, Zhang, & Nisbett, 2004), values (Ji, Nisbett, & Su, 2001; Kwan, Bond, & Singelis, 1997; Spencer-Rodgers, Williams, Hamilton, Peng, & Wang, 2007), preferences (Corriveau et al., 2017; DiYanni, Corriveau, Kurkul, Nasrini, & Nini, 2015; Liang & He, 2012) and self-concepts (Spencer-Rodgers, Boucher, Mori, Wang, & Peng, 2009; Spencer-Rodgers, Boucher, Peng, & Wang, 2009). As a result, the Western and East Asian cultures are increasingly treated as cultural poles in efforts to measure cultural differences (Muthukrishna et al., 2020) and to correct for the pervasive bias in psychology research toward US and European samples (Arnett, 2016; Henrich, Heine, & Norenzayan, 2010; Nielsen, Haun, Kärtner, & Legare, 2017). Despite a long empirical tradition of comparisons between these two cultures and an 55 abundance of psychological accounts for observed differences, estimates of differences are difficult to compare quantitatively because of the varying samples, measures, and methods used in different reports. Further, many of the most prominent reports of cross-cultural differences predate the field-wide discussion of methodological issues in psychology research during the past 10 years (Open Science Collaboration, 2015). For example, much research

in this tradition has been exploratory and hence has not followed current guidance

regarding limiting analytic flexibility in order to decrease false positives (Simmons, Nelson,

63 & Simonsohn, 2011). Given the importance of evidence about specific cross-cultural

64 differences for constructing theories of culture more broadly (e.g., Markus & Kitayama,

65 1992, 2010), further investigation of many empirical findings is likely warranted.

Some empirical evidence points to issues in the robustness of cross-cultural 66 measurements. Typically, measures used in this literature are not standardized and do not have published evidence about reliability and validity (Flake & Fried, 2020). The few extant direct comparisons between measures of cultural difference suggest that theoretically related tasks, such as implicit and explicit measures of the same construct, might not cohere (e.g., Kitayama, Park, Sevincer, Karasawa, & Uskul, 2009). Further, in a study with twenty cross-cultural measures used within a single US sample, Na et al. (2010) found a lack of coherence between tasks measuring social orientation and cognitive style, observing only 8 significant correlations between tasks across 90 statistical tests. Finally, more recent work failed to replicate cultural differences on several related measures (Mercier, 75 Yama, Kawasaki, Adachi, & Van der Henst, 2012; Mercier, Zhang, Qu, Lu, & Van der Henst, 2015; Zhou, Gotch, Zhou, & Liu, 2008). Thus, there is a need for exploration of the reliability of individual tasks as well as the intercorrelations between them. 78

Our goal in the current study was to collect a large dataset on a range of
cross-cultural measures that had previously been used in comparisons of East Asian and
Western cultures, enabling investigations of the robustness of these differences in new
samples of Chinese and US participants. We made the decision to pursue the strategy of
gathering relatively large and heterogeneous convenience samples using online recruitment,

¹ These authors interpreted their findings as imply the measures are orthogonal – indexing different constructs – and concluded that group-level differences between cultures are unlikely to relate to within-group individual differences. However an alternative possibility is that the reliabilities of many individual tasks are low, a feature which would ensure low correlations between them.

rather than recruiting smaller, more matched samples using in-lab recruitment. Our
reasoning was that the larger samples that we could access using online recruitment would
allow us to conduct highly-powered statistical tests, allowing us to make well-powered tests
for cultural differences. Further, larger samples would afford the analysis of individual and
demographic differences within culture, a topic of considerable interest in this literature
(Na et al., 2020; e.g., Na et al., 2010). Finally, the development of browser-based online
versions of prominent cross-cultural tasks would allow their inspection and reuse by other
researchers, thus promoting a more cumulative approach to the measurement of cultural
differences.

Our experiments were intended to be close replications of the original studies, but
differences in format of administration introduced inevitable differences, in some cases
more substantial than others. The interpretation of discrepant outcomes between an
original study and a replication is complex, given that disparate outcomes can occur for
many reasons (Machery, 2020; Nosek & Errington, 2020; Zwaan, Etz, Lucas, & Donnellan,
2018). In our case interpretation is especially difficult and we explicitly avoid interpreting
our results as bearing on the status of the original findings we investigate.

There were some significant differences between our experiments and the original 100 studies. First, we recruited online convenience samples from the U.S. and China. Previous 101 work varied in the country of origins of participants (in several cases, Japan for East Asian 102 participants; Canada for Western participants), largely focused on either college students 103 or community members, and were administered more than a decade ago. The within-culture variations and generational differences between our samples and previous samples make results difficult to compare directly. Furthermore, our strategy of 106 constructing a battery of replication studies and administering them uniformly online 107 significantly altered the contexts in which the participants engaged with the tasks as well 108 as in some cases requiring alterations to the task itself. 109

Thus, our replication studies should be taken as an assessment of robustness: in
particular, whether a set of previously-reported East-West cross-cultural differences can be
recovered in online convenience populations. They are not assessments of the veracity of
the original findings. Nevertheless, we believe that the field of cross-cultural psychology
can be advanced via the identification of tasks that yield cross-cultural differences robustly
across a variety of samples and administration formats – we hope our work contributes to
this aim. We return to these interpretive issues in the General Discussion.

Our task selection process was initially shaped by an interest in relational reasoning 117 and accounts explaining it with reference to cross-cultural differences in visual attention 118 and social cognition (Duffy, Toriyama, Itakura, & Kitayama, 2009; Kuwabara & Smith, 119 2012; Moriguchi, Evans, Hiraki, Itakura, & Lee, 2012). Additionally, in Experiment 1, we 120 selected tasks that could potentially be administered to young children as well as adults, 121 for use in future work addressing developmental questions about the relative time course of 122 cross-cultural differences across the visual, social, and cognitive domains. We balanced four 123 desiderata in our task selection, preferentially choosing tasks that (1) had been 124 theoretically or empirically implicated in relational reasoning, (2) were associated with 125 differential performance in US-China comparisons or related cultural contrasts (e.g., East Asian vs. Western cultures), (3) were relatively short, accessible tasks appropriate for web administration, and (4) were related to vision or social cognition accounts for relational 128 reasoning. We further conducted an extensive set of pilot tests to ensure that participants 129 understood instructions and that the tasks yielded interpretable data. 130

In Experiment 2, we selected a second set of tasks to investigate based in part on the results of Experiment 1. In particular, we repeated a handful of tasks from Experiment 1, in some cases, varying task parameters. We then selected a further set of tasks that probed both cross-cultural differences in higher-level cognition (e.g., language and reasoning) and perception, again respecting the desideratum that the tasks should be relatively short and amenable to administration in a web browser. The final set of tasks included in each

Experiment is listed in Table 1.

In addition to the goal of replicating individual tasks, our hope was that the 138 relatively large dataset that we collected could be used to explore the structure of within-139 and across-cultural variation in cognition and perception more broadly. Towards this goal, 140 we included a relatively extensive demographic questionnaire in both of our Experiments, 141 with the aim of using these measures to explore variation within our samples. In the final section of the paper, we report a series of exploratory analyses. The first of these assess the reliability of individual tasks, aiming to gauge whether individual tasks are reliable enough from a psychometric point of view to support further individual differences analyses. We then report across-task correlations, aiming to discover covariation between tasks that might indicate that they load on the same construct. Finally, we turn to analyses of whether within-culture demographic variables predict variation in task performance. 148 Overall, a number of tasks revealed acceptable levels of reliability, but tasks did not cluster 149 together and we found relatively few demographic predictors of within-culture variation. 150

We make all code and data from our experiments available for further data collection and analysis in hopes of promoting further cumulative work on measures and theories of cross-cultural variation.

Table 1
Tasks included in each experiment and the final sample size after exclusion.

Experiment	Task	Citation	Task Description	CN US
1	Ambiguous Relational Match- To-Sample	Carstensen et al. (2019)	Infer whether an object or relation is causally relevant	167 169
	(RMTS)		v	
	Picture Free Description	Imada, Carlson, & Itakura (2013)	Describe pictures from memory after a brief study period	167 169
	Ebbinghaus Illusion	Imada, Carlson, & Itakura (2013)	Judge the size of circles in a context designed to bias size judgments	167 169

	Horizon Collage	Senzaki, Masuda, & Nand (2014)	Make an image by dragging and drop- ping stickers onto a display	167 169
	Symbolic Self- Inflation (Fam- ily)	Kitayama et al. (2009)	Draw self and family members as circles	141 110
	Uniqueness Preference	Kim & Markus (1999)	Choose a sticker from five stickers, four of which are the same color	167 169
	Child Causal Attribution	Seiver, Gopnik, & Goodman (2013)	Watch short vignettes and explain the decisions of the characters	167 169
	Raven's Progressive Matrices	Su (2020)	Use analogical reasoning to complete visually-presented patterns	167 169
2	Ambiguous Relational Match-To-Sample (RMTS)	Carstensen et al. (2019)	Infer whether an object or relation is causally relevant	174 293
	Picture Free Description	Imada, Carlson, & Itakura (2013)	Describe pictures from memory after a brief study period	132 284
	Change Detection	Mausda & Nisbett (2007)	Find differences in the foreground or background of two images	160 253
	Symbolic Self-Inflation (Friends)	Kitayama et al. (2009)	Draw a sociogram with self and friends as nodes, relationships as edges	158 252
	Adult Causal Attribution	Morris & Peng (1994)	Read a crime story and explain the criminal's motiva- tions	114 293
	Taxonomic- Thematic Simi- larity	Ji, Zhang, & Nisbett (2004)	Match items based on taxonomic or thematic similarity (e.g., cow: chicken / grass)	178 295

Semantic Intu-	, , ,	Decide whether a	181	298
ition	& Snedeker (2018)	story refers to a		
		named character		
		(whose actions are		
		mischaracterized)		
		or the person who		
		performed the ac-		
		tions (but had a		
		different name)		
Raven's Progres-	Su (2020)	Use analogical rea-	181	298
sive Matrices		soning to complete		
		visually-presented		
		patterns		

Experiment 1

In Experiment 1, our goal was to evaluate cross-cultural differences in a variety of constructs. We assembled a web-based battery of tasks and tested these on a snowball sample of US and Chinese participants.

58 Methods

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Participants. We recruited participants through snowball sampling seeded at large universities in the US and China, in which participants directly recruited by the researchers were encouraged to recruit their friends and family members through email forwarding and social media sharing. Participants in the US were compensated with \$5 gift certificates (USD) and participants in China received ¥35 (CNY).

We recruited 203 and 201 participants each from the US and China, respectively.

Since we did not have strong a priori expectations about specific effect sizes, our overall

preregistered sample size was chosen to meet or exceed the sample sizes used in prior

reports in the literature from which our tasks were drawn.

Our original preregistered exclusion plan was to exclude people from the full dataset if they failed quality checks on any one task. However, due to a task demand associated

with the Symbolic Self-Inflation task, this criterion would have led to the exclusion of 85
people (US: 59, CN: 26) due to this task alone. As a result, we deviate from our
preregistration and include participants in the broader dataset even if they failed the
quality check for the Symbolic Self-Inflation task.

After exclusions, the US sample included 169 participants (44 Male, 114 Female, 9
Non-binary, 2 Declined to answer), with a mean age of 21.79 years old, all of whom were
native English speakers. The China sample included 167 participants (51 Male, 112
Female, 1 Non-binary, 3 Declined to answer), with a mean age of 22.49 years old, who were
all native speakers of Mandarin Chinese. This sample size is shared among all tasks except
for the Symbolic Self-inflation task, which included 110 US participants and 141 CN
participants.

In addition to age, gender and linguistic background, we collected a range of
demographic information including subjective socioeconomic status measured using the
MacArthur Ladder (Adler, Epel, Castellazzo, & Ickovics, 2000), level of maternal
education, the state or province the participant grew up in, residential mobility, and
number of overseas experiences.

Participants completed an online, browser-based sequence of eight Procedure. 186 tasks (see Table 1) and a brief demographic questionnaire. All tasks were implemented in a 187 combination of isPsych (De Leeuw, 2015) and custom HTML/JavaScript code. Tasks were 188 administered in English for the US sample and in Mandarin Chinese for the China sample. 189 To control for the impact of order-related inattention, task order was randomized across 190 participants with two exceptions: (1) the two drawing tasks (Symbolic Self-Inflation and 191 Horizon Collage) were always back-to-back in random order, and (2) Uniqueness Preference was always the penultimate task (in keeping with the task cover story, which congratulated 193 participants on being nearly done with the experiment). In total, the experiment took 194 about 30 minutes to complete. 195

Measures. Below, we give a short description of the methods for each task.

Ambiguous cRMTS. Carstensen et al. (2019) observed cross-culturally distinct developmental trajectories in a causal relational match-to-sample (cRMTS) task, and different preferences in an ambiguous formulation of this task. Specifically, when 3-year-olds saw evidence consistent with both object-based (e.g., blue cubes make a machine play music) and relational (pairs of different objects, AB, make a machine play music) solutions, children in the US sample preferentially chose the object-based solution, while those in China chose the relational solution.

We used an ambiguous version of the task (Carstensen et al., 2019, Experiment 3) to
explore whether adults in the US and China also show differing preferences for
object-based or relational solutions. Our participants saw two pairs of objects, AB and AC,
activate a machine, and were given a forced choice between an object-based solution (a
same pair of A objects, AA) and a relational solution (different pair BC).

Picture Free descriptions. Imada, Carlson, and Itakura (2013) found that 209 children around the age of 6 showed cultural differences in describing pictures to others. 210 Relative to US children, Japanese children tended to mention the objects in the 211 background first, as opposed to the focal objects in the picture. They also tended to 212 provide more descriptive accounts of the background objects than their US counterparts. 213 In our version of the task, we used a subset of seven images from the original study and adapted the task for adult participants, who studied each image for 5 seconds and then 215 typed a description. We coded the first mentioned item (focal or background) and counted 216 descriptors for focal and background elements. 217

Ebbinghaus Illusion. Both Japanese adults and children have been found to be
more susceptible to the Ebbinghaus Illusion – in which context alters the perceived size of
a circle – than Western participants in the US and UK (Doherty, Tsuji, & Phillips, 2008;
Imada et al., 2013). In this task, we followed the Imada et al. (2013) implementation of the

task, with two testing blocks: the No Context block (10 trials) and Illusion block (24 trials).

The No Context block establishes baseline accuracy for discriminating which of two orange circles is larger. In the Illusion trials, the two orange circles are flanked by a grid of 8 gray circles, which are all smaller or larger than the center circle. The illusion occurs because the orange circles appear larger when flanked by smaller gray circles, leading to distortions in comparing the sizes of the two orange circles with differing contexts (i.e., small or large flankers). Across the 24 Illusion trials, we measured accuracy of circle size judgments as a function of the actual size difference and flanker context (helpful or misleading).

Senzaki, Masuda, and Nand (2014) found that school-age Horizon Collage. 230 children in Japan and Canada showed culture-specific patterns when creating a collage of 231 an outdoor scene. Japanese children would draw the horizon higher and put more collage 232 items in the picture, relative to Canadian children. We adapted the task from Senzaki et 233 al. (2014) study 2, in which participants were prompted to make a collage with stickers. 234 Our participants could drag any of thirty images (line-drawings of people, animals, houses, 235 etc.) onto a rectangular "canvas" in the middle of the screen. There was also a sticker 236 "horizon," a horizontal line that spanned the length of the canvas. All stickers, including 237 the horizon, could be clicked and dragged to the canvas to produce "a picture of the 238 outside." Participants were asked to include a horizon and any number of other stickers to 239 create their image. We measured the height of the horizon, the number of stickers used, and the total area occupied by stickers (Senzaki et al., 2014). 241

Symbolic Self-Inflation. Kitayama et al. (2009) found a difference between
Western and East Asian cultures in the size of circles participants drew to represent
themselves relative to other people in their social networks. Japanese participants drew
circles of similar sizes to represent themselves and others, while those from Western
countries (US, UK, Germany) tended to draw their "self" circles larger than those
representing others, indicating a symbolic self-inflation in the three western cultures
compared to Japan. We adapted this task, asking participants to draw themselves and the

family members they grew up with as circles by clicking and dragging the mouse on a rectangular "canvas" to draw circles of varying sizes. They then labeled each circle for the person it represented. We measured the diameter of each circle and calculated a percent inflation score for each participant by dividing the diameter of the self circle by the average diameter of circles for all others.

Uniqueness Preference. Kim and Markus (1999) tested East Asians' and 254 Americans' preferences for harmony or uniqueness by asking them to pick one gift pen 255 from five options. In the condition that we replicated, the options differed only in the 256 barrel colors – four were the same and one was unique. They found that European 257 Americans were more likely to choose the unique colored one than East Asian participants. 258 We adapted our task to better fit the format of our online experiment by showing a virtual 250 "sticker book" to measure progress through all tasks in our study. At the end of each task, 260 participants received a virtual sticker. For the uniqueness preference task, we let them 261 select one of five dinosaur stickers: four blue dinosaurs and one yellow. Choice of the 262 unique vs. repeated color was recorded. 263

Causal Attribution. Previous work has shown that participants from South 264 Korea and the U.S. attribute behaviors differently in situations where there is evidence in 265 favor of situational explanations (Choi, Nisbett, & Norenzayan, 1999). Similarly, Chinese 266 media is more likely than U.S. media to attribute a person's behaviors to situational 267 context as opposed to individual traits (Morris, Nisbett, & Peng, 1995; Morris & Peng, 268 1994). We adapted the deterministic situation condition in Seiver, Gopnik, and Goodman 269 (2013), a task originally designed for children. In this task, two children both engage in one activity and avoid another, suggesting that situational constraints (e.g., the latter activity 271 being dangerous) may be guiding their decisions. Participants watched a series of four short, animated vignettes in which two children both played in a pool and neither child 273 played on a bicycle. We then asked participants to explain in text why each child did not 274 play on the bicycle, making for two test trials per participant. We used the prompt

question from Seiver et al. (2013), which explicitly pits person attributions against situational ones: "Why didn't Sally play on the bicycle? Is it because she's the kind of person who gets scared, or because the bicycle is dangerous to play on?" We coded each response for per-trial count of (a) personal and (b) situational attributions.

Raven's Standard Progressive Matrices. As an additional attention check as
well as an exploratory measure of relational reasoning assessing performance rather than
preference, we included the 12 questions from Set E of Raven's Standard Progressive
Matrices. Su (2020) found cross-cultural differences between adults in the US and China in
performance on this set. This set of questions was selected because it was the most difficult
subset and also the one most dependent on true analogical reasoning (without alternative
heuristic approaches like visual pattern completion).

Analytic approach. Our sample size, methods, and main analyses were
pre-registered and are available at https://aspredicted.org/37y6a.pdf. Data and analysis
scripts are available at FIXME

The specific papers that we drew on for our tasks used a heterogeneous set of analytic 290 methods. Rather than planning to replicate these specific analyses, we instead attempted 291 to follow current best practices by using linear mixed effects models with maximal random 292 effect structure as a unified analytic framework (Barr, Levy, Scheepers, & Tily, 2013). We 293 fit a separate model to each task. In case of convergence failure, we followed standard 294 operating procedure of pruning random slopes first and then random intercepts, always maintaining random intercepts by participant. For linear models, we report p-values derived from t-scores. For linear mixed models, we report p-values derived from z-scores, which is appropriate for relatively large samples (Blouin & Riopelle, 2004). Our key tests of interest were typically either the coefficient for a main effect of country (US/China) or 290 an interaction of country and condition. 300

Results

Ambiguous cRMTS. To examine whether adults in the US and China show differing preferences for object-based or relational solutions, we ran a mixed-effects logistic regression predicting response choice (object or relation) with country (US or China) as a fixed effect. There was no main effect of country on response choice (object or relation; US: M = 0.39, SD = 0.48; CN: M = 0.37, SD = 0.47; $\beta = 0.14$, SE = 0.89, z = 0.16, p = 0.87). The preference for object-based solutions seen in US preschoolers and the corresponding preference for relational solutions observed in China in an ambiguous context did not extend to adults in our samples.

Our US results replicate findings by Goddu and Walker (2018), who reported that US
adults are at chance in this paradigm. It seems likely that adults in both groups of our
study are aware of the ambiguous evidence and their near-chance selections reflect
(reasonable) uncertainty.

Picture Free Description. Based on Imada et al (2013), we expected Chinese participants would be more likely to mention background objects first and provide more descriptive accounts for background objects relative to focal objects, in comparison with US participants. Our results extend previous findings with the former metric (first mention; US: M = 0.90, SD = 0.17; CN: M = 0.56, SD = 0.30) but not the latter (number of descriptive accounts; For focal objects: US: M = 1.06, SD = 0.51; CN: M = 0.88, SD = 0.44; For background objects: US: M = 1.31, SD = 0.94; CN: M = 0.94, SD = 0.72).

For first mention, we ran a mixed-effects logistic regression predicting the type of first mention (object or relation) with country (US or China) as a fixed effect. We found a main effect of country ($\beta = 3.36$, SE = 0.34, z = 9.94, p < 0.01). For descriptive accounts, we ran a mixed-effect Poisson regression model predicting the number of descriptive accounts, with description type (focal or background), country (US or China), and their interaction as fixed effects. There was a significant main effect of culture (with US participants

providing more descriptions overall: $\beta = 0.36$, SE = 0.13, t = 2.68, p < 0.01). The culture effect interacted with the description types, but the effect was in the opposite direction, with U.S participants provided more background descriptions than focal descriptions, relative to Chinese participants ($\beta = -0.16$, SE = 0.07, t = -2.16, p < 0.05).

The mixed results between the first mention and descriptive accounts measures suggest that there is some complexity in linking broader theoretical accounts to specific measures; we interpret this result with caution and include the task in Experiment 2 to follow up further.

Ebbinghaus Illusion. To test whether perception of the Ebbinghaus illusion 335 varied across populations in our sample, we ran a mixed-effects logistic regression 336 predicting accuracy on each trial, with country (US or China), context (No Context or 337 Illusion Context), and circle size difference (the percent of difference in diameters) as fixed 338 effects, along with their interactions. We found main effects of context (with worse 339 performance in the Illusion Context; $\beta = 4.95$, SE = 0.29, z = 17.03, p < 0.01) and circle 340 size difference (worse performance for smaller differences; $\beta = 0.34$, SE = 0.01, z = 27.33, 341 p < 0.01). There was a marginally significant main effect of country at the opposite of the 342 predicted direction (US participants performed worse: $\beta = 0.52$, SE = 0.26, z = 1.95, p =343 0.05) but no interactions with country (All $\beta < 0.01$; All p > 0.05).

In sum, we failed to replicate cultural differences found between Western (US/UK) and Japanese participants in susceptibility to the Ebbinghaus illusion.

Horizon Collage. In the Horizon Collage task, three key measurements are
calculated from the "collage" participants created: the height of the horizon (height in
proportion to the height of the frame), the number of stickers, and the total area of the
stickers covered (following the original analysis, we added up the area occupied by each
individual sticker); Japanese children tend to put the horizon higher and include more
stickers that cover more area in their collage, compared with Canadian children. We ran a

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fixed effect linear model with culture as the main predictor for each of the measurements.

Culture did not significantly predict any of the three measurements (Sticker height: US: M = 0.57, SD = 0.15; CN: M = 0.54, SD = 0.20; Sticker number: US: M = 11.51, SD = 5.81;

CN: M = 11.77, SD = 5.80; Sticker area: US: M = 16.98, SD = 8.36; CN: M = 17.43, SD = 8.60; All $\beta < 0.03$; All p > 0.1).

Our experiment contrasted Chinese and US adults, rather than Japanese and
Canadian children. Although Senzaki et al. (2014) found that the cultural differences were
more salient in older children than younger children, suggesting that cultural differences
might increase with development, interpretation of our failure to replicate is still qualified
by differences in culture and medium of administration.

Symbolic Self-Inflation. To test whether US adults have a larger symbolic self

than Chinese adults, we ran a linear regression predicting percent inflation score (calculated 364 by dividing the diameter of the self circle by the average diameter of circles for others) with 365 country (US or China) as a fixed effect. No difference was found in the degree of symbolic 366 self-inflation between US and China adults based on percent inflation scores (US: M =367 0.95, SD = 0.26; CN: M = 0.95, SD = 0.55; $\beta = < 0.01$, SE = 0.06, t = 0.04, p = 0.97). 368 One possible explanation for our null results is that there are also cultural differences 369 between Japan and China in self concept; Japanese samples typically demonstrate 370 characteristics previously associated with East Asian cultures in general, with Chinese 371 samples deviating from these characteristics at times (Bailey, Chen, & Dou, 1997; Church 372 et al., 2012, 2014). In addition, the null results could also be attributed to our task design 373 being different from Kitayama et al. (2009). Instead of asking participants to draw their social network, our design asked participants to draw themselves and the family members 375 they grew up with. During the coding process, we noticed that people from both cultures 376 tended to draw older people, e.g., their parents, into larger circles, which might have 377 resulted in overall larger circles for other people than the self-circles in our task for both 378

cultures, masking any US-China difference in the degree of self-inflation. We follow up with

this possibility by changing the task prompt in Experiment 2.

Uniqueness Preference. We examined cross-cultural preferences for uniqueness 381 by running a simple logistic regression predicting each participant's single choice (minority 382 or majority color) with country (US or China) as a fixed effect; we used logistic regression 383 rather than mixed effects logistic regression due to the absence of repeated observations. 384 There was not a large cross-cultural difference in the probability of choosing the uniquely 385 colored sticker (US: M = 0.57, SD = 0.50; CN: M = 0.63, SD = 0.48; $\beta = -0.23$, SE = 0.48386 0.22, z = -1.02, p = 0.31). 387 The difference between our result and that of the original study by Kim and Markus 388 (1999) might be related to the use of online format in our study. In the original study, 389 participants were asked to pick a gift pen from five physical pens with different barrel 390 colors. It could be that Asian American participants in the previous study chose the more 391 common color because they wanted the next person to also have room for decision making 392 in the face of resource scarcity, or because they were expressing values or identities 393 influenced by East Asian cultural mandates favoring interpersonal harmony and similarity. 394 Our finding is also consistent with previous work demonstrating that tendencies toward conformity in East Asian samples are linked to reputation management (Yamagishi, Hashimoto, & Schug, 2008); it may be that our online experiment did not establish a 397 sufficient social context to motivate participants' concern about reputation, and accordingly failed to motivate reputation management in the form of a conformity preference. 399 Causal Attribution. To test whether Chinese participants tended to make more 400 situational attributions, and US adults more personal attributions, we ran a mixed-effects Poisson regression predicting the number of attributions included in each explanation, with attribution type (situational or personal), country (US or CN), and their interaction as fixed effects. We found a main effect of attribution type (Situational attribution: US: M =0.65, SD = 0.61; CN: M = 0.65, SD = 0.52; Personal attribution: US: M = 0.33, SD = 0.65405 0.55; CN: $M=0.51,\,SD=0.52;\,\beta=0.24,\,SE=0.10,\,z=2.37,\,p<0.05).$ Neither the

interaction nor the main effect of culture was significant (Both $\beta < 0.3$; p > 0.05).

The failure to find cross-cultural differences in attributions could be related to the style of the tasks, which was relatively repetitive and originally designed for children; in Experiment 2, we follow up with a causal attribution task designed for adults.

Raven's Standard Progressive Matrices. As an exploratory measure of relational reasoning, we ran a mixed-effects logistic regression predicting per-trial accuracy, with country as a fixed effect, random intercepts for each subject and question, and by-question random slopes for country. We found a main effect of country, with Chinese participants outperforming those from the US (US: M = 0.68, SD = 0.24; CN: M = 0.84, SD = 0.17; $\beta = -1.31$, SE = 0.23, z = -5.64, p < 0.01).

Our findings replicate Su (2020) in finding an advantage for Chinese participants on Raven's Matrices. In our context, we also interpret the relatively high scores we observed as evidence that participants were engaging fully with our tasks.

Discussion

We did not observe cross-cultural differences in the majority of the tasks in 421 Experiment 1. The only exceptions were in picture description and our exploratory 422 measure of reasoning performance (Raven's Matrices). Many of our tasks did not have a 423 manipulation check and could yield null results simply by virtue of inattention. However, 424 the results of the Raven's task (and the Ebbinghaus Illusion) suggest that participants 425 were engaged in our tasks and performed at a high objective level. Further, in addition to minor methodological changes that we made, interpretation of our failures to replicate individual tasks in many cases could be due to (1) differences in administration (online vs. in-person), (2) differences in participant recruitment (e.g., university pool vs. snowball 429 recruitment), (3) differences in target age (adults vs. children), and (4) differences in 430 sample (e.g. Japanese vs. Chinese adults in the East Asian group). 431

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Our failure to find robust Western vs. East Asian cultural differences in this initial
selection of tasks was dispiriting. We designed Experiment 2 to extend Experiment 1 by
recruiting a different sample and identifying followup or replacement tasks that we hoped
would yield a broader set of cross-cultural differences.

Experiment 2

Experiment 2 was designed to follow up on Experiment 1 and further evaluate

cross-cultural differences across a battery of tasks. Since several of our tasks in Experiment 438 1 yielded no evidence for cross-cultural differences, we replaced these with alternative tasks 439 selected to address similar or related constructs. We replaced the Ebbinghaus Illusion with a measure of Change Detection that had been argued to index context sensitivity (Masuda & Nisbett, 2006). We replaced the child-appropriate causal attribution task with a task designed for adults (Morris & Peng, 1994). We also included two tasks measuring linguistic or semantic intuitions more broadly (Taxonomic/Thematic Similarity and Semantic Intuition), following up on the detection of cross-cultural differences in the Picture Free 445 Description task. Although our goal in Experiment 2 was to evaluate a further set of tasks, 446 we also included the RMTS, Picture Free Description, and Raven's Progressive Matrices 447 tasks to replicate our results from Experiment 1, and we included a modified version of 448 Symbolic Self-Inflation to address several issues with the earlier version of the task. 449 In Experiment 2, we made use of crowd-sourcing services – rather than snowball 450 sampling – as our participant recruitment channel. We had two rationales. First, in 451 Experiment 1 our samples were quite young (due to the use of email and social media to populations of university students for recruitment). A younger sample might be more 453 exposed to international media and influences and be less likely to show distinct cross-cultural differences. Second, we were concerned that being recruited by friends and 455 family (as in a snowball sample) might prime interdependent thinking among our 456 participants, leading to decreased cross-cultural differences. 457

$_{458}$ Methods

Participants. We recruited participants through online crowdsourcing websites.

For the US, we used Prolific and applied the following screening criteria: a) U.S.

nationality; b) born in the U.S. and c) currently reside in the U.S.; For China, we used

Naodao (www.naodao.com), a platform designed for conducting online experiments in

mainland China. Participants in U.S. were compensated at the rate of \$12.25 per

submission and in China ¥35 per submission. We recruited 304 participants from the U.S.

and 185 participants from China.

10 participants were excluded because they did not meet our demographic inclusion

10 participants were excluded because they did not meet our demographic inclusion
criteria. Following our preregistration (available at https://osf.io/u7mzg), we applied a
task-based exclusion procedure in which we excluded a participant's responses in a
particular task if they a) showed a response bias in the tasks, b) had missing data on more
than 25% of trials or c) failed to meet the inclusion criteria for any specific task as specified
in the preregistration.

Similar to Experiment 1, we collected demographic information from participants, including subjective socioeconomic status, the state or province the participant grew up in and the one they currently reside in, residential mobility, number of international experiences, education, and undergraduate area of study (STEM or non-STEM). We also administered scales to collect explicit measures of participants' cultural identities and behaviors (Cleveland & Laroche, 2007; Cleveland, Laroche, & Takahashi, 2015).

The sample size for each task after exclusion and the descriptive statistics for each demographic question are reported in Table 1.

Procedure. Similar to Experiment 1, participants completed eight tasks and a
brief demographics questionnaire online. The experiment was administered online in
English for the US sample and in Mandarin Chinese for the Chinese sample, with the
exception of the Adult Causal Attribution task. The Adult Causal Attribution task was

administered in English, and only Chinese participants who self-identified as being able to 484 read English participated in this task. To control for the impact of order-related 485 inattention, task order was randomized across participants with two exceptions: (1) the 486 Free Description task always occurred before (not necessarily immediately) Change 487 Detection (because change detection includes a manipulation check that explicitly asks 488 about focal objects, which could bias responding in Free Description), and (2) the two 480 story-based tasks (Semantic Intuition and Adult Causal Attribution) always occurred 490 together in a fixed order at the end of the study, with Semantic Intuition first and Adult 491 Causal Attribution last. Adult Causal Attribution was always the last task (if run) because 492 it was administered in English and we did not wish to prime CN participants with English 493 stimuli before any of the other tasks, all of which were run in Mandarin.

Measures.

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Tasks repeated from Experiment 1. We replicated three tasks from
Experiment 1 using identical procedures: Ambiguous RMTS, Picture Free Description, and
Raven's Progressive Matrices.

Symbolic Self-Inflation. Participants were asked to draw themselves and their friends as circles, as opposed to drawing themselves and their family members as circles in Experiment 1. They were also asked to draw lines between any two people who are friends, as in the original study by Kitayama et al. (2009). They then labeled each circle to indicate the person it represents. We calculated a percent inflation score for each participant by dividing the diameter of the self circle by the average diameter of circles for others.

Adult Causal Attribution. We speculated that the lack of cultural difference in Causal Attribution in Experiment 1 might be due to the simplistic nature of our task, which was designed for use with young children. Therefore, in Experiment 2 we used a paradigm designed for adults, in which participants were asked to read a crime narrative from a news report that included substantial information on a criminal's background and

the events leading up to their crime, and then rate the relevance of various situational and 510 personal factors (Morris & Peng, 1994). In the original study, both Chinese participants 511 and US participants read stories in English. We followed this procedure by selecting the 512 subset of our Chinese participants who self-identified as comfortable reading short stories 513 in English to participate. In the task, participants were told that they would be reading 514 news stories and answering questions to help social scientists understand the factors that 515 contribute to murders. Participants were randomly assigned to read one of two stories 516 (Iowa shooting or Royal Oak shooting). After the stories, they were asked to write a short 517 explanation for the murderer's behaviors. Then, they rated a list of statements on a 518 7-point Likert scale about the extent to which each was a likely cause of the murder. The 519 statements included items that describe personal and situational factors. We measured 520 endorsement of these two factor types.

Masuda and Nisbett (2006) found differences in attention Change Detection. 522 allocation between Japanese and US participants in a change detection paradigm. They 523 found that Japanese participants were significantly faster than US participants in 524 identifying changes in the background of images. We followed their original procedure and 525 used the same stimuli. In this task, participants were presented with 30 pairs of images. 526 On each trial, two pictures would alternate on the screen, each presented for 560ms with a 527 blank screen in between images for 80ms. The two pictures were almost identical with 528 subtle differences, either in the focal object (e.g., a tractor in daylight with its lights on or 529 off) or the background (e.g., the change in a cloud's location in the background sky). 530 Participants were instructed to press a key when they spotted the difference, and then describe the difference in a text box. If they did not detect a difference within 60 seconds, the trial timed out. Only trials in which participants correctly identified the changes were included in the analysis. After 30 trials, participants saw each pair of images again, this 534 time side-by-side on the screen. They were asked to identify the focal object(s) in the 535 pictures by typing into a text box. These responses were used as a manipulation check to

ensure that participants in both cultures construed focal objects similarly.

We coded difference descriptions to exclude trials in which participants did not identify the change, and checked agreement on focal objects across cultures. We measured how quickly participants identified the difference on trials in which they reported the difference correctly.

Taxonomic/thematic similarity task. Ji et al. (2004) showed that Chinese 542 participants are more likely to categorize items based on thematic similarity, whereas US participants are more likely to categorize items based on taxonomic similarity. In this task, participants were presented with a list of word sets. Each set contained three words: a target word as prompt and two other words as options. The list included test sets and filler sets. In each test set, one option was a taxonomic match (e.g. monkey - elephant) and the 547 other a thematic match (e.g. monkey - banana). In each filler set, the cue item and the 548 options were broadly similar, thematically and taxonomically, making for a more 549 ambiguous decision (e.g. monkey - elephant, tiger). Participants completed a 550 two-alternative forced choice task (2AFC) in which they chose one match for each cue item. 551

Ji et al. (2004) is replicated in a more recent work (Le, Frank, & Carstensen, 2021), so
we used a subset of testing materials from Le et al. (2021), including 15 test triads, 15 filler
triads, and 2 attention check questions. The order of the triads was randomized between
subjects. We measured taxonomic vs. thematic match selections on each of the test trials.

Semantic Intuition. Li, Liu, Chalmers, and Snedeker (2018) found cultural
differences in semantic intuitions about ambiguous referents in Chinese and US
participants. Chinese participants are more likely to determine the referent of a name
based on the description of the speaker (the descriptivist view) whereas U.S. participants
are more likely to determine the referent based on the original usage (the causal-historical
view). In the study, participants read five separate stories and judged the correctness of
statements referring to a character after each story. Two comprehension check questions

were included for each story. We followed the original testing procedure closely and used
the same materials. We measured participants' semantic intuition as their judgment on the
correctness of statements referring to the critical characters.

666 Results

583

Ambiguous RMTS. Our analysis was identical to that in Experiment 1. We did 567 not observe a main effect of country on participants' preference for object vs relational 568 matches (Proportion of relational match: US: M = 0.41, SD = 0.44; CN: M = 0.41, SD = 0.41569 0.42; $\beta = -0.01$, SE = 0.48, z = -0.03, p = 0.98). As in Experiment 1, we did not find 570 evidence that the differential preferences observed in preschoolers extend to adults. It seems 571 likely that adults in both populations are aware of the mixed evidence for the relational 572 and object solution and that their responses reflect sensitivity to this ambiguous design.² 573 **Picture Free Description.** US participants were more likely to mention the focal 574 objects than the background objects (First mention: US: M = 0.94, SD = 0.14; CN: M =0.69, SD = 0.26). We used the same regression analysis as in Experiment 1 and found a 576 main effect of country ($\beta = 3.09$, SE = 0.32, z = 9.61, p < 0.01). Our results replicate Experiment 1's finding for the first-mention measure with comparable effect size (standardized mean difference; Experiment 1: 1.48[1.24, 1.72]; Experiment 2: 1.57[1.34, 579 1.80]). ³. These findings extend Imada et al.'s (2013) findings to Chinese adults. 580 Change Detection. We ran a linear mixed-effects model predicting the reaction 581 time to correctly identify changes in the pictures, with country (U.S. or China) and type of 582

change detected (focal or background) as main effects, as well as their interaction. We did

² Our reliability analysis shows that adults expressed this uncertainty only at the population level: individuals tended to be consistent in choosing the same solution type across all four test trials, with ambiguity expressed as disagreement between participants.

³ The comparable SMD suggests that the finding was not caused by the idiosyncrasy of our samples. As a result, we decided not to code the descriptive accounts for Experiment 2 that did not show differences in Experiment 1.

not find evidence for an interaction between culture and type of change detected ($\beta = 0.04$, 584 SE = 0.03, z = 1.40, p = 0.16). Participants in both countries identified changes to the 585 context faster than changes to focal objects (Context Changes: M = 10,101.87, SD =586 4,257.15; Focal Object Changes: M = 10,646.54, SD = 4,816.10; $\beta = 0.07$, SE = 0.02, t = 0.02587 3.45, p < 0.01). Chinese participants identified both types of change more quickly than 588 U.S. participants (US: M = 10,689.49, SD = 4,406.73; CN: M = 9,875.67, SD = 4,733.57; 580 $\beta = 0.12$, SE = 0.05, t = 2.27, p < 0.05). In sum, we did not replicate the findings of 590 Masuda and Nisbett (2006). 591

Symbolic Self-Inflation. In Experiment 1, we did not find a significant difference in the degree of symbolic self-inflation between adults in the US and China. Here, we observed a pattern contrary to the prediction: U.S. adults showed less self-inflation than Chinese adults (US: M = 1.30, SD = 0.51; CN: M = 1.45, SD = 0.65; $\beta = -0.15$, SE = 0.06, t = -2.56, p < 0.05). In sum, we did not replicate the findings of Kitayama et al. (2009) (with Japanese participants) in either of our experiments.

Adult Causal Attribution. We ran a mixed-effects linear regression predicting 598 endorsement of each potential cause with country (U.S. or China) and attribution type 599 (personal or situational) as fixed effects, as well as their interaction. We found an 600 interaction in the predicted direction: Chinese participants endorsed situational 601 attributions to a greater extent than their counterparts in the U.S. (Situational ratings: US: M = 1.71, SD = 0.80; CN: M = 3.17, SD = 0.89; Personal ratings: US: M = 3.12, SD= 1.10; CN: M = 3.14, SD = 1.07; β = -1.39, SE = 0.14, t = -9.71, p < 0.01). This result 604 extends the original findings by Morris and Peng (1994), and suggests that the measure of 605 causal attribution in Experiment 1 (which was designed for use with child participants) 606 may not be appropriate for measuring cross-cultural differences in causal attribution 607 among adults. 608

Taxonomic-Thematic Similarity. We used a mixed-effects logistic regression
model predicting response (taxonomic or thematic match) with country (US or China) as a

fixed effect. There was a significant effect in the predicted direction: participants in the U.S. were more likely to choose taxonomic matches than participants in China (Proportion of taxonomic matches: US: M=0.69; SD=0.46; CN: M=0.44; SD=0.50; $\beta=2.02$, SE=0.89, t=2.27, p<0.05). This finding replicates the findings of Ji et al. (2004) and Le et al. (2021).

Semantic Intuition. We ran a mixed-effects logistic regression predicting response (descriptive or causal-historical) with country (US or China) as a fixed effect, and found that U.S. participants made significantly more causal-historical choices than Chinese participants (Proportion of causal historical choice: US: M = 0.71; SD = 0.46; CN: M = 0.53; SD = 0.50; $\beta = 1.59$, SE = 0.37, t = 4.37, p < 0.01). We also replicated the item effect identified by Li et al. (2018), though this was not among our preregistered analyses. In sum, We replicated Li et al. (2018) with a new sample of US and China adults.

Raven's Standard Progressive Matrices. We replicated the findings from Experiment 1. Chinese participants scored higher on Raven's Standard Progressive Matrices than U.S. participants (US: M=0.49, SD=0.27; CN: M=0.73, SD=0.23; β = -1.82, SE=0.25, z=-7.39, p<0.01).

7 Discussion

Overall, Experiment 2 was more successful than Experiment 1, with more cultural
differences found between the U.S. and Chinese participants. The success can be attributed
to the inclusion of the successful tasks from Experiment 1 (e.g. Free Description and
Ravens Progressive Matrices), and the exclusion of tasks designed for young children
(e.g. Child Causal Attribution, Horizon Collage).

Exploratory analysis

We conducted a set of exploratory analyses to consolidate results from the two
experiments. We first performed a mini-meta analysis with the tasks from both
experiments. Then, we assessed the reliability of the tasks that included multiple trials.
Finally, we examined how explicit cultural identities and demographic factors relate to task
performance.

639 Mini-meta analysis

As our first exploratory analysis, we identified the key effect of interest from our
pre-registration (usually a main effect of culture or an interaction of culture, depending on
task) and converted the coefficient into a standardized measure of effect size (standardized
mean difference; SMD) via the method described by Westfall, Kenny, and Judd (2014).
Because there is no "correct" direction for all of the tasks except Raven's Matrices, we
show the absolute value of effect size (Figure 1).

Across our two experiments, we saw consistent and generally large differences (SMD 646 > 0.6) in Free Description, Raven's Matrices, Adult Causal Attribution, Semantic Intuition 647 and Triads tasks. Aside from Raven's Matrices, all of these tasks had in common that they 648 were deliberative linguistic tasks that tapped into relatively high-level cognitive constructs. 649 In contrast, we observed effect sizes close to zero for our more aesthetic and perceptual 650 tasks (Change Detection, Ebbinghaus Illusion, and Horizon). We also observed little 651 consistent difference in four other tasks (RMTS, Symbolic Self-Inflation, Uniqueness 652 Preference, and Children's Causal Attribution), perhaps for reasons idiosyncratic to each. 653 We return to the broader question of generalization across task types in the General Discussion. 655

We next conducted a set of exploratory analyses to consolidate results from the two experiments. First, we assessed the reliability of the tasks that included multiple trials. We

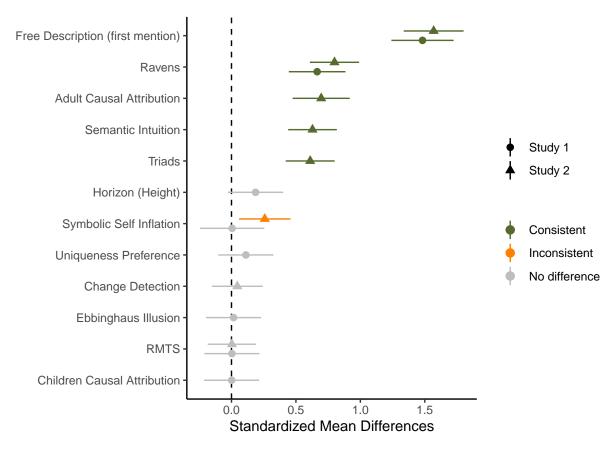


Figure 1. Forest plot of effect sizes (standardized mean difference) for each task across both experiments. Point shape shows experiment number and color provides a guide to whether effects were consistent with prior literature.

next examined whether there was shared variance between tasks. Finally, we examined how explicit cultural identities and demographic factors related to task performance.

660 Reliability assessment

One question motivating our work was whether the individual tasks we used were reliable enough – had low enough measurement error – to be used for further investigation of individual differences. The gold standard for the measurement of whether a task yields stable within-person measurements is test-retest reliability (simply because test-retest gives a direct estimate of stability over time), but this method was outside the scope of our study. Thus we used a split-half approach, asking whether participants' answers on individual questions related to one another. We used a permutation-based split half

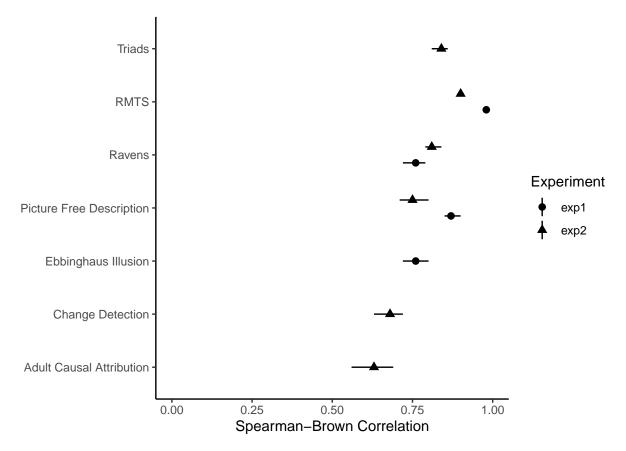


Figure 2. Spearman-Brown adjusted reliabilities for tasks with more than four trials. Point shape shows experiment number. Error bars show 95% confidence intervals.

approach (Parsons, 2021) in which we made 5000 random splits of items into two simulated

"halves" and then computed the within-person correlation between scores on these two

halves, averaging across simulated runs. To estimate the reliability of the full-length

instrument, we used the Spearman-Brown "prophecy" formula.

Since split-half approach is only suitable for tasks with multiple trials, we removed
tasks with less than four trials from the analysis. For tasks with more than one condition,
we focused on the conditions that were predicted to show cultural differences (i.e. Illusion
context condition for Ebbinghaus task; Situational judgement for Adult Causal Attribution
task; Context condition for Change Detection task).

Figure 2 shows the corrected split-half reliabilities for all tasks in both of our
experiments. Overall, the reliabilities were acceptable (all Spearman-Brown Correlations >

o.6). We further investigated whether there was cultural variation in the reliability of tasks. For most tasks, the reliabilities were relatively similar (within 0.1 of one another), but there were three tasks where reliability was lower for US participants than Chinese participants: Change Detection (US - CN = -0.19), Adult Causal Attribution (US - CN = -0.31), Free Description in Study 1 (US - CN = -0.23).

Relations between individual tasks

One (perhaps simplistic) interpretation of the prior literature on cultural variation is 685 that there is a general tendency toward holistic or analytic reasoning that varies across 686 cultures and explains variation in tasks. This single dimension might correspond to broad 687 (or focused) attention and contextualized, relational reasoning (or an emphasis on focal 688 people or objects). As a first step towards investigating this interpretation, we explored 689 whether there was a single dimension of individual variation in our data that corresponded 690 to the general axis of cross-cultural difference. Because some data was missing, largely due 691 to task-related exclusions, we treated the missing data using two approaches: listwise deletion and imputation with means. These approaches yielded comparable results, so here we report correlations from listwise deletion. 694

Correlations between task scores were quite low on average, suggesting limited support for the hypothesis of a single factor explanation. Across both Experiments 1 and 2, the largest absolute magnitude of correlations observed were -0.29 (Triads and Adult Causal Attribution in study 2) and -0.28 (Free Description and Ravens in Study 2), and -0.24 (Adult Causal Attribution and Free Description in Study 2). All other correlations were between -0.23 and 0.23. Hence, the amount of shared variation between tasks was quite limited and our attempts at exploratory factor analysis discovered structures with many distinct factors and very low loading on the first factor.

Demographic variation and explicit measures of cultural identity

As a final exploratory analysis, we asked whether demographic variation or variation 704 in cultural identity predicted responding in our tasks. Our approach to these questions was 705 to fit a set of exploratory regression models for each task, predicting task scores as a 706 function of an individual scale and its interaction with culture. This approach allowed us 707 to explore both within- and across-culture effects in a single model. Our predictors were 1) 708 the summed score for our global/local cultural identity and consumption measures (with local items reverse-scored, such that higher scores represent more global identity and 710 consumption patterns), 2) geographic information about where participants grew up Markus & Conner (2014), and 3) a range of demographic factors, including age, gender 712 identities, residential mobility, number of international experiences, maternal education 713 level, and subjective socioeconomic status as measured by the MacArthur Ladder (Adler et 714 al., 2000). 715

Task – Global identity relationships. We fit models predicting task scores
based on culture and its interaction with global – local identity for tasks in Experiment 2
(we did not collect these scales in Experiment 1). Two of these relationships were
statistically significant at .01 p = 0.05; Triads task: p = 0.04) but neither of these relationships survived Bonferroni correction for multiple
comparisons across the family of coefficients for the models across all tasks.

Task – Geographic origin relationships. We next considered whether regions within each country were meaningful predictors of task performance. We fit models predicting task scores based on the categories of regions the participants reported grew up in. For China, provinces were categorized as rice-cultivating regions and wheat-cultivating regions based on Talhelm et al. (2014). For U.S., states were categorized based on either the coastal locations (West Coast, East Coast and Inland) or broad geographic locations (West, South, Northeast, Midwest), following the categorization reported in Carstensen,

Saponaro, Frank, and Walker (2022).

5 out of 48 models we ran showed a statistically significant relationships between 730 regions and task performance. In Study 1, coastal location was a significant predictor for 731 Free Description task. Participants who grew up in Inland regions or East Coast were more 732 likely to mention the focal object first when describing the pictures (Inland: p = 0.02; East 733 Coast: p = 0.05). In Study 2, both coastal location and broad geographic location were 734 significant predictors for Ravens, with participants from Inland and East Coast scoring 735 higher than participants from West Coast (Inland: p < 0.01; East Coast: p = 0.05), and 736 participants from Midwest and South scoring higher than participants from the West 737 (Midwest: p < 0.01; South: p = 0.04). In addition, the two categories also predicted 738 performance in Change Detection. East Coast participants took longer to respond than 739 West Coast participants (p = 0.02), and Northeastern participants took longer to respond than participants grew up in the West (p < 0.01). However, none of these relationship survived Bonferroni correction.

Basic demographic effects. We fit 192 exploratory regression models to see if
basic demographic factors could predict task performance. The demographic factors we
explored were age, gender identities, residential mobility, number of international
experiences, maternal education level, subjective socioeconomic status as measured by
MacArthur Ladder (Adler et al., 2000). 24 were statistically significant, but only one model
survived Bonferroni correction. Change detection was predicted by age in the U.S. sample,
with older participants taking longer to respond than younger participants (Adjusted p < 0.01).

General Discussion

The world's cultures are strikingly different, and psychologists have long sought to measure and characterize this variation, with differences between Western and East Asian cultures as a particular case study of interest. These efforts have given rise to a rich literature documenting cultural differences in a wide range of psychological tasks. Across
two experiments, we selected a range of tasks that had previously been shown to yield
differences between Western and East Asian samples and replicated them with two
relatively large online samples of US and Chinese participants. In this discussion, we first
consider the limitations of our study since these contextualize the remainder of our
conclusions. Next we consider the interpretation of our results within individual tasks. We
end the discussion with a summary of the key findings of this work.

762 General Limitations

As discussed above and in the introduction, we did not design our experiments to replicate prior work directly, and hence one important limitation of our work is simply that it cannot be used as a test of the reliability of prior findings. Instead, our measures provide estimates of US-China differences on a range of constructs, specifically for online convenience samples. These estimates are likely biased downward – towards the null hypothesis of no difference between cultures – by several features of our experimental design.

Online experiments (especially grouped into a long battery as ours were) likely
receive slightly less attention than in-person studies, though overall these effects have
tended to be small in US samples (Buhrmester, Kwang, & Gosling, 2016). Contra this
concern, however, participants did perform relatively accurately on those tasks that had
correct answers (e.g., Raven's Matrices, Ebbinghaus Illusion), and in our exploratory
analysis, we found relatively high reliabilities on all tasks. Further, our pre-registered
exclusion criteria removed participants who performed poorly. Thus, we do not believe that
participants were inattentive overall.

Another limitation of our estimates of US-China differences comes from differences in sampling strategy between cultures. In Experiment 1, we used the same snowball sampling

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procedure, but this procedure may have yielded different samples due to differences in 780 social networks or norms about sharing study information across cultures. In Experiment 781 2, because the platform we used to recruit U.S. participants (Prolific) was not accessible in 782 China, we used a different platform to recruit Chinese participants (Naodao). Prolific and 783 Naodao have different levels of popularity and different participant pools, resulting in some 784 asymmetry between the US and Chinese samples. Despite these differences between 785 samples both across and within experiments, we do not see indications that our estimates 786 were dramatically biased by our sampling decisions. First, our results were largely 787 comparable in the tasks that were included in both experiments (e.g. Picture Free 788 Description; Ravens; and RMTS). Second, in our exploratory analyses we did not find 789 strong associations between participant demographics and cross-cultural effects (with some 790 small exceptions discussed in that section). Finally, we reran all of our preregistered analyses with an age-matched subset of U.S. participants in Experiment 2 and found our results were qualitatively identical.

Thus, while our samples are certainly not representative samples of US or Chinese
national populations – indeed to our knowledge, nearly all work to date has used
convenience samples of one type or another – they appear to yield stable cross-sample
estimates that do not reflect large biases due to sampling strategy or demographics.

One of the main ways in which our samples may not have been representative is that
they are likely to be more globalized than the population on average simply by being young
(and thus less acculturated) and having access to a computer. Contra this concern,
variation in local cultural identity did not strongly relate to variation in any of our tasks,
but interestingly, we observed the strongest local identities (within our Chinese sample)
among the youngest participants.

Last but not least, another difference between our experiments and previous work was the lack of an experimenter, and some of our tasks may be particularly sensitive to the

presence of an experimenter. In a web experiment, participants are often isolated in front of their own computer. In contrast, when participating in an in-person experiment, participants need to interact with and perform the task in front of the experimenters who are often from the same social group. Indeed, in the uniqueness preference pen choice task, cross-cultural differences are dependent on the presence of an experimenter (Yamagishi et al., 2008). Our null results, obtained in the absence of an experimenter, can be seen as a conceptual replication of this work.

813 Task-specific Limitations

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In addition to the general limitations discussed above, there are features of our experimental adaptations that may have affected performance in specific tasks. In this section, we highlight concerns about these issues and discuss their implications for interpreting the results of these tasks. See Table 2 for summary of task-specific discussion.

In the case of the Uniqueness Preference task, it is possible that adapting the task to 818 an online format in which resource scarcity was not strictly real and choices in this task 819 had no lasting effect (in the form of a new pen), may have trivialized the choice and 820 undermined the incentive for prosocial, harmonious behavior or expression. This possibility 821 is consistent with the chance responding we observed in both groups. Alternatively, our 822 results could be seen as a conceptual replication of Yamagishi et al. (2008), who argue that 823 differences in this task are moderated by the likelihood of evaluation, with no differences in 824 pen choice observed in the absence of an experimenter. 825

The ambiguous developmental tasks, Ambiguous RMTS and Child Causal
Attribution, may have been too heavy-handed in their key manipulations; both were
designed to highlight ambiguity for young children, but it may be that their explicit cues
and repetitive instructions impressed this ambiguity too strongly for adult audiences,
resulting in the adults' near-chance responding – a reasonable response to such marked

ambiguity. Cultural differences in causal reasoning and attribution may only manifest
when the task design is age-appropriate. Consistent with this view, we did replicate
previously attested differences in the Adult Causal Attribution task in Experiment 2, and
other recent work has shown cross-cultural differences in causal attribution among 4- to
9-year-olds in Germany, Japan, and Ecuador using a design similar to the Child Causal
Attribution task (Jurkat, Iza Simba, Hernández Chacón, Itakura, & Kärtner, 2022).

Last but not least, cultural variation within the broader constructs of East Asia and 837 the West could explain some of our findings as a failure to extend previous work. Some of the tasks we included originally compared children and adults from other parts of East 839 Asia and the West (e.g., Horizon Collage, Symbolic Self Inflation, Change Detection; but c.f. Masuda, Ishii, and Kimura (2016) for an alternative account of mixed findings in change detection paradigms). For example, the Taxonomic-Thematic Similarity task 842 replicated previously attested cross-cultural differences between the US and China both 843 here and in other work (Le et al., 2021) but these differences failed to generalize to a 844 US-Vietnam comparison, despite the cultural, historical, and geographic similarities 845 between China and Vietnam. This variation suggests that similar psychological tendencies 846 could be expressed differently under distinct sociocultural contexts and traditions, even 847 across regions and countries that share many similarities. As another example, responding 848 in the Horizon Collage task could be modulated by variation between countries: Chinese 849 and Japanese aesthetic traditions differ, so while Chinese and Japanese people may share a 850 preference for highly contextualized information, this preference may typically be expressed 851 through distinct visual techniques. 852

3 Conclusion

We conducted two sets of experiments to examine the robustness of several classic
experimental paradigms in cross-cultural psychology. Our results showed a heterogeneous
pattern of successes and failures: some tasks yielded robust cultural differences across both

experiments, while others showed no difference between cultures. We estimated the
reliability of the tasks to be moderate, with only minor cultural variations. In addition, we
also explored the effect of a range of demographic variables, including explicit identification
with global identity, regional differences within cultures, and several demographic
characteristics. All of these had minimal relation to task performance.

Our goal here was not to perform direct replications that would shed light on the 862 replicability of specific findings. Instead, since our methods, administration medium, 863 sample, and analytic approach differed from the prior literature, our hope was to examine 864 the robustness of these paradigms as a method for measuring US-China differences in an 865 online context. Our work has several strengths relative to the prior literature, including larger samples of participants from the US and China, two broad groups of tasks 867 implemented openly online (and reusable by future researchers), and a preregistered 868 analysis plan that allows for the unbiased estimation of cross-cultural effects. In sum, we 869 hope that our work here provides a foundation for future studies that seek to establish a 870 robust and replicable science of cross-cultural difference. 871

Tasks included in each experiment and the final sample size after exclusion. Table 2

Task Name o Ambiguous Relational Match-To- Sample (RMTS)	 Age: The original experiment was conducted with adults. Test Format: The original experiment was conducted in person with physical stimuli and live experimenters; the current experiment was contacted in person with physical stimuli and live tention to the calculative reasons Age: The relation be have been less sational experiment was contacted in person with physical stimuli and live tention to the calculative reasons 	Speculative reasons The relation between objects have been less salient when the were presented as schematic ani gaged with the tasks and paid tention to the causal propertie
	ducted through web-based interfaces with animation.	object pairs.

Ebbinghaus Illu-

sion

- Age: The original study was conducted with young children; the current experiment was conducted with adults.
- Culture: The original experiment compared Japanese participants with Canadian participants; the current experiment compared Chinese participants with U.S. participants."

- ts might he stimuli nimation.
- ll less end less aties of the
- Adults in both cultures reached ceiling performance.
- different levels of visual context sensitivity compared to Canadian and U.S. and Chinese participants may Japanese participants. •

Horizon Collage

- Age: The original experiment was conducted with young children; the current experiment was conducted with adults.
- Test Format: The original experiment was conducted in person with canvas and collage stickers; the current experiment was conducted through web-based interfaces.
- Culture: The original experiment compared Japanese participants with Canadian participants; the current experiment compared Chinese participants with U.S. participants.
- Test Format: The original experiment was conducted in person with pen and paper; the current experiment was conducted through web-based interfaces.

Self-(Fam-

Symbolic Inflation ily version and

Friends version)

Test Prompt: The original experiment asked the participants to draw their close social network; the current experiment asked the participants to draw their family members or friends growing up.

- The task might have been too trivial for adults to engage with properly.
- The saliency of horizon height is diminished by the drag-and-drop online interface.

The online interface might have altered participants' drawing process, making it more difficult to implicitly represent the symbolic meaning of the circle size.

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Test Format: The original experiment was conducted in person; the current experiment was conducted through web-based interfaces.

sents is less meaningful than a choice

for a real pen.

The choice for a virtual sticker repre-

Stimulus: The original experiment was giving away physical pens; the current experiment was asking the participants to select virtual stickers.

Child Causal Attribution

Age: The original experiment was conducted with young children; the current experiment was conducted with adults.

Change

Detection

Test Format: The original experiment was con-

ducted in person; the current experiment was con-

ducted through web-based interfaces.

The presence of a live experimenter might increase the social pressure on participants, causing them to consider the cultural perception of their choice. Adults may have failed to engage with their causal reasoning because the

story is too simplistic.

- The in-lab setting might facilitate participants' performance by making them pay more attention to the computer screen.

References

- Adler, N. E., Epel, E. S., Castellazzo, G., & Ickovics, J. R. (2000). Relationship of subjective and objective social status with psychological and physiological functioning. *Health Psychology*, 19(6), 586.
- Arnett, J. J. (2016). The neglected 95%: Why american psychology needs to become less american.
- Bailey, J. R., Chen, C. C., & Dou, S.-G. (1997). Conceptions of self and
 performance-related feedback in the US, japan and china. *J Int Bus Stud*, 28(3),
 605–625.
- Barr, D. J., Levy, R., Scheepers, C., & Tily, H. J. (2013). Random effects structure for confirmatory hypothesis testing: Keep it maximal. *Journal of Memory and Language*, 68(3), 255–278.
- Blouin, D. C., & Riopelle, A. J. (2004). The difference between t and z and the difference it makes. *J of Gen Psych*, 131(1), 77–84.
- Buhrmester, M., Kwang, T., & Gosling, S. D. (2016). Amazon's mechanical turk: A
 new source of inexpensive, yet high-quality data?
- Carstensen, A., Saponaro, C., Frank, M. C., & Walker, C. M. (2022). Bridging

 cultural and cognitive perspectives on similarity reasoning. In *Proceedings of the*annual meeting of the cognitive science society (Vol. 44).
- Carstensen, A., Zhang, J., Heyman, G. D., Fu, G., Lee, K., & Walker, C. M. (2019).

 Context shapes early diversity in abstract thought. *PNAS*, 116(28),

 13891–13896.
- Chan, C. C., Tardif, T., Chen, J., Pulverman, R. B., Zhu, L., & Meng, X. (2011).

 English-and chinese-learning infants map novel labels to objects and actions

 differently. *Dev Psy*, 47(5), 1459.
- Cheng, L. (2020). The development of cognitive styles among american and chinese children (PhD thesis).

910

911

912

- Choi, I., Nisbett, R. E., & Norenzayan, A. (1999). Causal attribution across cultures. *Psy Bull*, 125(1), 47.
- Chua, H. F., Boland, J. E., & Nisbett, R. E. (2005). Cultural variation in eye movements during scene perception. *PNAS*, 102(35), 12629–12633.
- Church, A. T., Alvarez, J. M., Katigbak, M. S., Mastor, K. A., Cabrera, H. F.,
 Tanaka-Matsumi, J., et al.others. (2012). Self-concept consistency and
 short-term stability in eight cultures. *J Res Pers*, 46(5), 556–570.
- Church, A. T., Katigbak, M. S., Ibáñez-Reyes, J., Jesús Vargas-Flores, J. de, Curtis,
 G. J., Tanaka-Matsumi, J., et al.others. (2014). Relating self-concept
 consistency to hedonic and eudaimonic well-being in eight cultures. *J Cross Cult*Psy, 45(5), 695–712.
 - Cleveland, M., & Laroche, M. (2007). Acculturation to the global consumer culture:

 Scale development and research paradigm. *Journal of Business Research*, 60(3),
 249–259.
- Cleveland, M., Laroche, M., & Takahashi, I. (2015). The intersection of global consumer culture and national identity and the effect on japanese consumer behavior. *Journal of International Consumer Marketing*, 27(5), 364–387.
- Corriveau, K. H., DiYanni, C. J., Clegg, J. M., Min, G., Chin, J., & Nasrini, J. (2017). Cultural differences in the imitation and transmission of inefficient actions. *J Exp Child Psy*, 161, 1–18.
- De Leeuw, J. R. (2015). jsPsych: A JavaScript library for creating behavioral experiments in a web browser. *Behavior Research Methods*, 47(1), 1–12.
- DiYanni, C. J., Corriveau, K. H., Kurkul, K., Nasrini, J., & Nini, D. (2015). The role of consensus and culture in children's imitation of inefficient actions. *J Exp Child Psy*, 137, 99–110.
- Doherty, M. J., Tsuji, H., & Phillips, W. A. (2008). The context sensitivity of visual size perception varies across cultures. *Perception*, 37(9), 1426–1433.

- Duffy, S., Toriyama, R., Itakura, S., & Kitayama, S. (2009). Development of cultural strategies of attention in North American and Japanese children. *J Exp Child Psy*, 102(3), 351–359.
- Flake, J. K., & Fried, E. I. (2020). Measurement schmeasurement: Questionable measurement practices and how to avoid them. Advances in Methods and Practices in Psychological Science, 3(4), 456–465.
- Goddu, M., & Walker, C. M. (2018). Toddlers and adults simultaneously track
 multiple hypotheses in a causal learning task. In *CogSci*.
- Henrich, J., Heine, S. J., & Norenzayan, A. (2010). The weirdest people in the world? *Behav Brain Sci*, 33 (2-3), 61–83.
- Imada, T., Carlson, S. M., & Itakura, S. (2013). East—west cultural differences in context-sensitivity are evident in early childhood. *Dev Sci*, 16(2), 198–208.
- Ji, L.-J., Nisbett, R. E., & Su, Y. (2001). Culture, change, and prediction. *Psych*Sci, 12(6), 450–456.
- Ji, L.-J., Peng, K., & Nisbett, R. E. (2000). Culture, control, and perception of relationships in the environment. *JPSP*, 78(5), 943.
- Ji, L.-J., Zhang, Z., & Nisbett, R. E. (2004). Is it culture or is it language? *JPSP*, 87(1), 57.
- Jurkat, S., Iza Simba, N. B., Hernández Chacón, L., Itakura, S., & Kärtner, J.

 (2022). Cultural similarities and differences in explaining others' behavior in

 4-to 9-year-old children from three cultural contexts. *Journal of Cross-Cultural Psychology*, 00220221221098423.
- Kim, H., & Markus, H. R. (1999). Deviance or uniqueness, harmony or conformity?

 A cultural analysis. *JPSP*, 77(4), 785.
- Kitayama, S., Park, H., Sevincer, A. T., Karasawa, M., & Uskul, A. K. (2009). A cultural task analysis of implicit independence. *JPSP*, 97(2), 236.
- Kuwabara, M., & Smith, L. B. (2012). Cross-cultural differences in cognitive

- development: Attention to relations and objects. *J Exp Child Psy*, 113(1), 20–35.
- Kwan, V. S., Bond, M. H., & Singelis, T. M. (1997). Pancultural explanations for life satisfaction: Adding relationship harmony to self-esteem. *JPSP*, 73(5), 1038.
- Le, K., Frank, M., & Carstensen, Alex. (2021). Is it language or is it culture?

 Re-examining cross-cultural similarity judgments using lexical co-occurrence

 (Undergraduate Honors Theses). Stanford University.
- Li, J., Liu, L., Chalmers, E., & Snedeker, J. (2018). What is in a name?: The
 development of cross-cultural differences in referential intuitions. *Cognition*, 171,
 108–111.
- Liang, B., & He, Y. (2012). The effect of culture on consumer choice: The need for conformity vs. The need for uniqueness. *Int J of Consum Stu*, 36(3), 352–359.
- Machery, E. (2020). What is a replication? *Philosophy of Science*, 87(4), 545–567.
- Markus, H. R., & Conner, A. (2014). Clash!: How to thrive in a multicultural world.

 Penguin.
- Markus, H. R., & Kitayama, S. (1992). The what, why and how of cultural psychology: A review of shweder's thinking through cultures. *Psychological Inquiry*, 3(4), 357–364.
- Markus, H. R., & Kitayama, S. (2010). Cultures and selves: A cycle of mutual constitution. *Perspectives on Psychological Science*, 5(4), 420–430.
- Masuda, T., Ishii, K., & Kimura, J. (2016). When does the culturally dominant
 mode of attention appear or disappear? Comparing patterns of eye movement
 during the visual flicker task between european canadians and japanese. *Journal*of Cross-Cultural Psychology, 47(7), 997–1014.
- 977 Masuda, T., & Nisbett, R. E. (2006). Culture and change blindness. *Cognitive* 978 *Science*, 30(2), 381–399.
- Mercier, H., Yama, H., Kawasaki, Y., Adachi, K., & Van der Henst, J.-B. (2012). Is

1006

- the use of averaging in advice taking modulated by culture? J Cog & Culture, 12(1-2), 1-16.
- Mercier, H., Zhang, J., Qu, Y., Lu, P., & Van der Henst, J.-B. (2015). Do easterners and westerners treat contradiction differently? *J Cog & Culture*, 15(1-2), 45–63.
- Moriguchi, Y., Evans, A. D., Hiraki, K., Itakura, S., & Lee, K. (2012). Cultural differences in the development of cognitive shifting. *J Exp Child Psy*, 111(2), 156–163.
- Morris, M. W., Nisbett, R. E., & Peng, K. (1995). Causal attribution across domains and cultures.
- Morris, M. W., & Peng, K. (1994). Culture and cause: American and chinese attributions for social and physical events. *JPSP*, 67(6), 949.
- Muthukrishna, M., Bell, A. V., Henrich, J., Curtin, C. M., Gedranovich, A.,

 McInerney, J., & Thue, B. (2020). Beyond WEIRD psychology: Measuring and

 mapping scales of cultural and psychological distance. *Psych Sci*, 31(6), 678–701.
- Na, J., Grossmann, I., Varnum, M. E., Karasawa, M., Cho, Y., Kitayama, S., & Nisbett, R. E. (2020). Culture and personality revisited: Behavioral profiles and within-person stability in interdependent (vs. Independent) social orientation and holistic (vs. Analytic) cognitive style. *Journal of Personality*, 88(5), 908–924.
- Na, J., Grossmann, I., Varnum, M. E., Kitayama, S., Gonzalez, R., & Nisbett, R. E.

 (2010). Cultural differences are not always reducible to individual differences.

 PNAS, 107(14), 6192–6197.
- Nielsen, M., Haun, D., Kärtner, J., & Legare, C. H. (2017). The persistent sampling bias in developmental psychology: A call to action. *J Exp Child Psy*, 162, 31–38.
- Nosek, B. A., & Errington, T. M. (2020). What is replication? *PLoS Biology*, 18(3), e3000691.
 - Open Science Collaboration. (2015). Estimating the reproducibility of psychological

- science. Science, 349(6251).
- Parsons, S. (2021). Splithalf: Robust estimates of split half reliability. *Journal of Open Source Software*, 6(60), 3041.
- Richland, L. E., Chan, T.-K., Morrison, R. G., & Au, T. K.-F. (2010). Young
 children's analogical reasoning across cultures: Similarities and differences. *J*Exp. Child. Psy., 105 (1-2), 146–153.
- Sabbagh, M. A., Xu, F., Carlson, S. M., Moses, L. J., & Lee, K. (2006). The development of executive functioning and theory of mind: A comparison of chinese and US preschoolers. *Psych Sci*, 17(1), 74–81.
- Seiver, E., Gopnik, A., & Goodman, N. D. (2013). Did she jump because she was
 the big sister or because the trampoline was safe? *Child Dev*, 84 (2), 443–454.
- Senzaki, S., Masuda, T., & Nand, K. (2014). Holistic versus analytic expressions in artworks: Cross-cultural differences and similarities in drawings and collages by

 Canadian and Japanese school-age children. *J Cross Cult Psy*, 45(8), 1297–1316.
- Simmons, J. P., Nelson, L. D., & Simonsohn, U. (2011). False-positive psychology.

 *Psych Sci, 22(11), 1359–1366.**
- Spencer-Rodgers, J., Boucher, H. C., Mori, S. C., Wang, L., & Peng, K. (2009). The dialectical self-concept. *Pers Soc Psy B*, 35(1), 29–44.
- Spencer-Rodgers, J., Boucher, H. C., Peng, K., & Wang, L. (2009). Cultural differences in self-verification. *J Exp Soc Psy*, 45(4), 860–866.
- Spencer-Rodgers, J., Williams, M. J., Hamilton, D. L., Peng, K., & Wang, L.

 (2007). Culture and group perception: Dispositional and stereotypic inferences
 about novel and national groups. *JPSP*, 93(4), 525.
- Su, S. (2020). Analogical reasoning in Chinese and US adults (Master's thesis).

 Cornell University.
- Talhelm, T., Zhang, X., Oishi, S., Shimin, C., Duan, D., Lan, X., & Kitayama, S. (2014). Large-scale psychological differences within China explained by rice

1045

1046

1047

1048

1049

- versus wheat agriculture. Science, 344 (6184), 603–608. 1034 Tan, B. (2020). Chinese and US young children's executive function and its 1035 sociocultural antecedents (PhD thesis). The University of Memphis. 1036 Tardif, T. (1996). Nouns are not always learned before verbs: Evidence from 1037 mandarin speakers' early vocabularies. Dev Psy, 32(3), 492. 1038 Waxman, S. R., Fu, X., Ferguson, B., Geraghty, K., Leddon, E., Liang, J., & Zhao, 1039 M.-F. (2016). How early is infants' attention to objects and actions shaped by 1040 culture? New evidence from 24-month-olds raised in the US and china. Front 1041 Psy, 7, 97. 1042 Westfall, J., Kenny, D. A., & Judd, C. M. (2014). Statistical power and optimal 1043 design in experiments in which samples of participants respond to samples of 1044
 - design in experiments in which samples of participants respond to samples of stimuli. Journal of Experimental Psychology: General, 143(5), 2020.

 Yamagishi, T., Hashimoto, H., & Schug, J. (2008). Preferences versus strategies as
 - explanations for culture-specific behavior. Psychological Science, 19(6), 579–584.
 - Zhou, J., Gotch, C., Zhou, Y., & Liu, Z. (2008). Perceiving an object in its context—is the context cultural or perceptual? *Journal of Vision*, 8(12), 2–2.
- Zwaan, R. A., Etz, A., Lucas, R. E., & Donnellan, M. B. (2018). Making replication mainstream. *Behavioral and Brain Sciences*, 41.