
Twin-2K-500: A dataset for building digital twins of over 2,000 people based on their answers to over 500 questions

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Paper under double-blind review

Abstract

1 LLM-based digital twin simulation, where large language models are used
2 to emulate individual human behavior, holds great promise for research
3 in business, AI, social science, and digital experimentation. However,
4 progress in this area has been hindered by the scarcity of real, individual-
5 level datasets that are both large and publicly available. To address this gap,
6 we introduce a large-scale, public dataset designed to capture a rich and
7 holistic view of individual human behavior. We survey a representative
8 sample of $N = 2,058$ participants (average 2.42 hours per person) in the
9 US across four waves with over 500 questions in total, covering a com-
10 prehensive battery of demographic, psychological, economic, personality,
11 and cognitive measures, as well as replications of behavioral economics
12 experiments and a pricing survey. The final wave repeats tasks from earlier
13 waves to establish a test-retest accuracy baseline. Initial analyses suggest
14 the data are of high quality and show promise for constructing digital twins
15 that predict human behavior well at the individual and aggregate levels.
16 Beyond LLM applications, due to its unique breadth and scale the dataset
17 also enables broad social science and business research, including studies
18 of cross-construct correlations and heterogeneous treatment effects.

19 1 Introduction

20 The rise of large language models (LLMs) like GPT has sparked interest across disciplines
21 (including marketing, computer science, economics, psychology and political science) in
22 leveraging these tools to create “silicon samples” which may replicate how these humans
23 would behave in response to any stimuli (Arora et al. 2024, Argyle et al. 2023, Brand et al.
24 2023, Dillion et al. 2023, Goli and Singh 2023, Horton 2023, Li et al. 2025, Park et al. 2023,
25 Qin et al. 2024). If these LLM-simulations can be a faithful substitute for eliciting responses
26 from their human counterparts, the implications for both academics and practitioners are
27 substantial. Academics could use silicon samples for pilot experiments to pinpoint stimuli
28 with significant impact, thus improving the efficiency of theory development and experi-
29 mental design. Firms could leverage these realistic simulations to explore different ideas
30 and strategies, thereby improving customer insight and product development. Accordingly,
31 in the recent past we have witnessed a large influx of firms offering services leveraging
32 silicon samples for customer insights (e.g., Synthetic Users, Outset AI, Nexxt, Voxpopme,
33 Evidenza, Expected Parrot, Meaningful, xPolls, Ipsos, CivicSync).

34 While silicon samples may be generated using only demographic information or hypo-
35 thetical “life stories,” a promising approach consists in creating silicon samples that are
36 “digital twins” of real people. Notably, Park et al. (2024) use LLMs to create digital twins of
37 over 1,000 individuals based on transcripts from qualitative interviews, and find that the
38 simulated agents replicated the human participants’ responses on the General Social Survey
39 85% as accurately as participants replicate their own answers two weeks later.

40 Despite the promise and excitement surrounding digital twins, some uncertainty remains.
41 For example, Brucks and Toubia (2025) show that the answers provided by LLMs may be
42 overly influenced by the architecture of the prompt, such as the labeling or ordering of
43 options in multiple choice questions. Gui and Toubia (2023) show that leveraging LLMs to

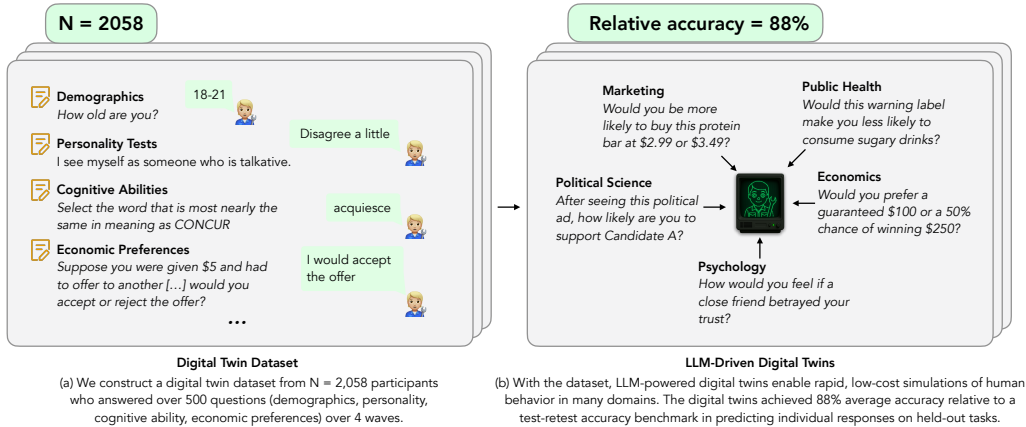
simulate experiments may introduce unwanted confounding, due to the difficulty of clearly instructing the LLM how to draw variables not specified in the prompt. Other research (Santurkar et al. 2023, Motoki et al. 2024, Li et al. 2025) suggests LLMs tend to express opinions that are not representative of the (human) population.

Given this background, it is crucial for the academic and practitioner community to validate digital twins in a transparent, reliable, and replicable manner. However, existing datasets present significant limitations that hinder their effectiveness for this purpose. Some datasets are publicly available (e.g., Alattar et al. 2018, Center 2023, Santurkar et al. 2023), but they are not well suited for testing the validity of digital twins because they do not contain behavioral data (e.g., experiments) nor a test-retest accuracy benchmark. Other datasets have this feature, but they are not publicly available (e.g., Park et al. 2024).

In sum, to the best of our knowledge there is no publicly available dataset that combines rich psychological profiles, behavioral data, and demographics from a large, representative sample for the development and testing of digital twin simulations. As a result, researchers often rely on synthetic or proprietary data, which undermines transparency, reliability, and replicability.

To address this gap, we assemble and publicly share an extensive dataset from a representative sample of $N = 2,058$ people who each answered over 500 questions covering a wide range of demographic questions, psychological scales, cognitive performance questions, economic preferences questions, as well as replications of a wide range of within- and between-subject experiments on heuristics and biases taken from the behavioral economics literature. The data was collected across 4 waves of studies lasting on average 2.42 hour per participant in total. Table 3 gives an overview of the measures collected in each wave and Figure 1 illustrates our overall approach.

Figure 1: Overview



We use the responses to the heuristics and biases questions from waves 1-3 as holdout data, and train the digital twins based on the rest of the data from waves 1-3. Wave 4 repeated the heuristics and biases experiments, providing us with a measure of test-retest accuracy. Future uses of the data may keep the same split, or combine all the data from the 4 waves to create digital twins.

We report encouraging results regarding the quality of the data: correlations between measures have good face validity, we replicate almost all known results from the behavioral economics literature, and the test-retest accuracy is robust.

We also report initial tests of the predictive validity of digital twins constructed using the data. At the individual level, we compute the accuracy of the digital twin predictions on holdout questions, against the test-retest accuracy benchmark as well as a random benchmark. At the aggregate level, we test whether the digital twin simulations replicate

the average treatment effects observed in human data. Throughout this process, we explore the type of behavior that can be predicted with higher vs. lower accuracy by the digital twins, to develop insight into the range of potential applications.

The dataset and code are publicly available.¹ Given the unique scale and breadth of the data, there is also value in the raw results, irrespective of the application to digital twins. We report descriptive statistics and correlations between the dozens of measures we collected. We encourage others to explore heterogeneous treatment effects (Dean and Ortoleva 2019, Stanovich and West 2008).

2 Methods

We assembled a wide-range of measures proposed in the business and social science literatures over the past several decades. In addition to 14 demographic questions, we included 19 personality tests that measured 26 constructs over 279 questions, 11 cognitive ability tests (85 questions, 11 measures), 10 economic preferences tests (34 questions, 10 measures). We also replicated 11 between-subject experiments (16 questions) and 5 within-subject experiments (32 questions) from the behavioral economics literature.² Finally, we administered the pricing study from Gui and Toubia (2023), which asks participants to make purchase decisions about 40 different products at randomly selected prices. In total, participants answered 500 questions across the first three waves. Wave 4 repeated all within- and between-subject heuristics and biases experiments from the first three waves as well as the pricing study from wave 3 (16+32+40=88 questions in total). Participants were assigned to the exact same condition in wave 4 as they were in waves 1-3 for each of these experiments, providing us with a clean measure of test-retest accuracy. We programmed the studies on Qualtrics, doing our best to replicate the stimuli and measures from the original papers.³

We launched wave 1 on Prolific on 01/29/2025, targeting 2,500 representative US respondents (sampled by age, sex, and ethnicity). Participants received \$7 for completing wave 1.⁴ They were informed that this was the first of four waves and they would earn a \$10 bonus for completing all waves (with comprehension checks to ensure understanding). We received 2,509 complete responses. The following week (02/04/2025), we invited these 2,509 participants to wave 2 (\$7), receiving 2,263 complete responses. The next week (02/11/2025), we invited them again for wave 3 (\$7), receiving 2,252 complete responses. Waves 2 and 3 were closed the next week. On 02/25/2025, we invited the 2,154 participants who had completed waves 1–3 to wave 4 (\$6; two-week delay since wave 4 repeated previous measures), receiving 2,058 complete responses and closing the wave after one week. Those completing all four waves received an additional \$10 bonus, totaling \$37.

These 2,058 participants who completed all four waves constitute our final sample. Among our final sample, the average response time was 43.88 minutes for wave 1 (std=19.26), 45.31 minutes for wave 2 (std=19.24), 32.66 minutes for wave 3 (std=15.68), and 24.09 minutes for wave 4 (std=12.51). The average total time across all 4 waves was 145.47 minutes (std=56.10).

¹The dataset is publicly available and the link is hidden for anonymization purpose, and the LLM simulation code can be found at <https://anonymous.4open.science/r/Digital-Twin-Simulation-04DF/>. (A non-anonymized github repository will be linked upon acceptance).

²We included all experiments in Stanovich and West (2008) who study “some of the most classic and well studied biases in the heuristics and biases literature,” as well as false consensus which allowed us to both capture participant’s opinions on a range of issues and to test another well-known bias.

³We made minor adjustments to reflect cultural and societal changes (e.g., in the mental accounting scenarios from Thaler (1985) we replaced “Mr. A” and “Mr. B” with “Person A” and “Person B,” and in the sunk cost experiment of Stanovich and West (2008) we replaced video rental stores with coffee shops).

⁴We pre-tested each wave to estimate response time and adjusted compensation accordingly.

3 Data

Section 5 presents initial results of the performance of digital twins created with the data. Those results are subject to change as researchers explore optimal ways to create digital twins, and the current results may be viewed as providing a lower bound on performance. In the current section, we instead focus on exploring the *intrinsic* quality of the data.

Table A1 reports demographic characteristics of our sample. While our digital twins are created based on the raw responses, it is also informative and potentially useful to extract the measures corresponding to these questions (e.g., the extraversion score is measured by averaging 8 questions from the Big Five battery of questions). Across waves 1-3, we collected 47 measures capturing personality traits, cognitive abilities and economic preferences. Studying the correlations between these measures is of general interest to business and social science scholars and practitioners, above and beyond the question of digital twins. Web appendix A3 details the construction of these measures from the raw data, and Table A2 reports summary statistics for the individual-level measures collected in the study. We compute a total of 1,326 pairwise correlations between the 47 measures listed in Table 3 and 5 demographic characteristics. We apply the Bonferroni correction and consider a correlation as significant if the p -value is below $\frac{0.05}{1326}$. This gives us 509 pairs of measures with significant correlations. We cannot report them all here, and instead report in Table A3 10 examples of correlations that are particularly high and/or noteworthy. These correlations all have good face validity, which suggests the data are of good quality, despite the large number of questions.

Next, we test whether our 16 heuristics and biases experiments replicate known results at the aggregate level. See Table 2. We see that both in waves 1-3 and in wave 4, all between-subject results replicate those in the literature, with the exception of the base rate fallacy. While Kahneman and Tversky (1973) find that probability assessments are not sensitive to base rate, we find that they are. In terms of within-subject experiments, waves 1-3 and wave 4 also replicate all known results, with the exception of the non-separability of risks and benefits for one of the items, bicycles. While Stanovich and West (2008) find a negative correlation between judged benefits and risks for this item, we find no correlation. The fact that our data replicates the vast majority of these known experimental results is again a sign of good data quality.

Finally, we calculate the test-retest accuracy in our data. We use the answers from waves 1-3 to our 88 holdout questions (across 17 tasks) as the ground truth. Given that all holdout questions are either binary or numerical (or transformed into numerical answers), we calculate accuracy as follows. For binary measures, accuracy is simply a binary indicator of whether two answers match. For non-binary measures, we calculate the absolute deviation between the ground truth and predicted answer, divided by the range of possible answers.⁵ We then compute accuracy as 1 minus this absolute deviation. This measure generalizes accuracy from binary to numerical questions: it ranges between 0 and 1, is equal to 1 when the prediction is equal to the ground truth, and 0 when it is maximally different. When multiple questions are included in the same task, we take the mean accuracy across the questions within each task. Therefore, we are left with one measure of accuracy per respondent for each of the 17 tasks (11 between-subject experiments, 5 within-subject experiments, 1 pricing study). Figure 2 reports the average accuracy across respondents for each task as well as 95% confidence interval. We see that the average test-retest accuracy across the 17 tasks is 81.72%. This number is aligned with others reported in the literature (e.g., Park et al. 2024), and again gives us confidence in the data's quality.

4 Creation of the digital twins

To construct each digital twin, we begin by merging the original Qualtrics survey files (QSF) with each participant's raw responses, creating a self-contained JSON record for every

⁵For the anchoring questions which accept unbounded answers, we transform the data into deciles based on the answers from wave 2 before calculating the absolute deviation.

individual. This record lists, in order, every question the participant actually encountered, the response options shown, and the answers. We then partition this record into three separate files:

- **Persona JSON:** Aggregates all non–hold-out content from waves 1-3, used to define the persona.
- **Evaluation answer-block JSON:** Contains the participant’s wave 1–3 responses to hold-out items, providing the ground truth for evaluating simulation accuracy.
- **Retest answer-block JSON:** Stores wave 4 responses to those same hold-out items, used solely to compute the human test–retest accuracy benchmark.

By distributing the data in this modular format, we enable future researchers to experiment with alternative encoding or summarization schemes before presenting the material to an LLM. Future research may also explore alternative ways to split the questions into training and hold-out observations, e.g., predicting the responses to cognitive ability questions based on the answers to the other questions in waves 1-3.

In the present work, we use a straightforward text-based approach: the JSON files are converted into text descriptions detailing the questions, options, and participant answers. The prompt instruction is attached in Web Appendix A2.1. The model’s completion is then post-processed back into canonical survey coding and compared with the Wave 1–3 ground truth, yielding the accuracy statistics reported in Figure 2. Our dataset includes all these files for future researchers to train and test LLM-based digital twin simulations.

5 Initial tests of digital twins’ predictive performance

Section 3 presented evidence that suggests the quality of the data is good, despite the high number of questions. Hence, this dataset should be helpful to researchers and practitioners interested in developing, testing, improving and deploying digital twin simulations. In this section, we present initial tests of the performance of digital twins created using the data, both at the individual and aggregate levels. As mentioned, those initial results may be viewed as providing a lower bound on the predictive performance that may be achieved in the future.

5.1 Individual level

To systematically evaluate different strategies for LLM-based persona simulation, we experimented with over a dozen variations in both persona construction and simulation methodology. These include differences in input format (e.g., text vs. JSON), model choices, prompting strategies such as proactive reasoning or chain-of-thought, and persona summaries. Full experimental details are provided in Web Appendix A2.⁶ Overall, we find that the predictive accuracy of the answers simulated by the digital twins falls within a similar range across approaches (see Table 1). We hope this collection of baseline results will serve as a useful benchmark for future researchers exploring more advanced methods for persona training, such as reinforcement learning with human feedback (RLHF). For the initial analyses reported here, we focus on the text format using GPT4.1-mini.

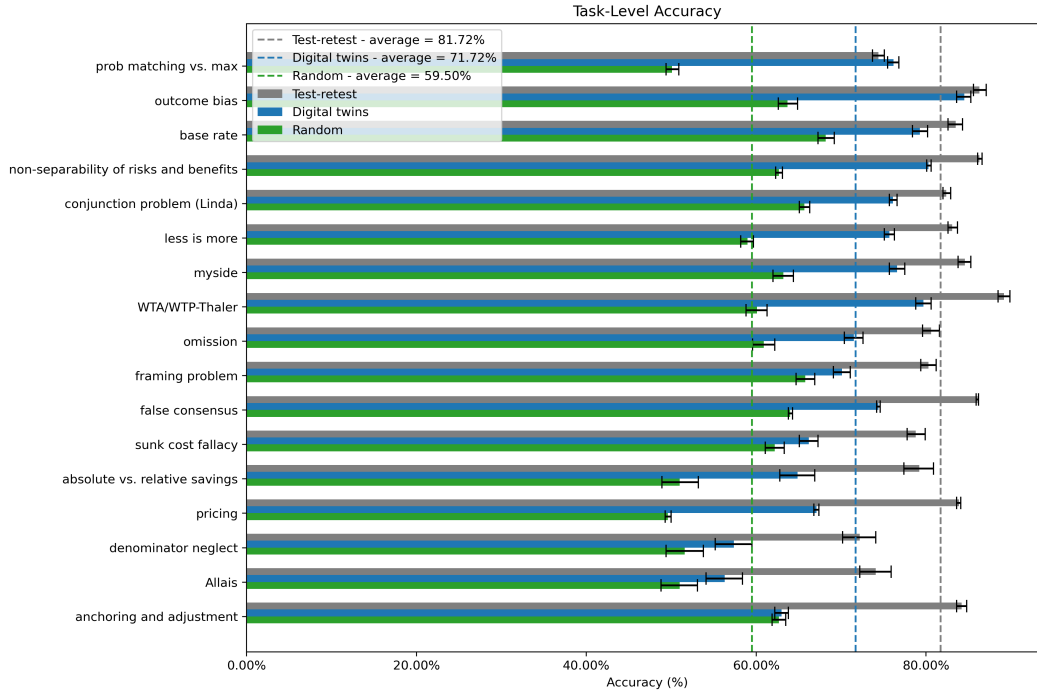
Figure 2 reports, for each task, the predictive accuracy of the answers simulated by the digital twins, as well as the accuracy of a random benchmark which chooses each answer from a random uniform distribution. On average, across the 17 tasks the accuracy of the digital twin predictions is 71.72%, and the ratio of the digital twin accuracy to the test-retest accuracy is 87.67%.

⁶The appendix also examines benchmarks that require some of the holdout data (split into a training and validation subsample), i.e., fine-tuning and a traditional machine learning benchmark XG Boost.

Table 1: Various persona simulation approaches and evaluation results

Approach	LLM	Accuracy
Text Persona	GPT4.1-mini	71.72%
Text Persona	Gemini-flash2.5	69.40%
JSON Persona	GPT4.1-mini	70.48%
JSON Persona	GPT4.1	71.05%
Persona Summary	GPT4.1-mini	68.02%
Persona Summary + JSON Persona	GPT4.1-mini	67.88%
Text Persona (Reasoning)	GPT4.1-mini	70.39%
Text Persona (Repeating Questions)	GPT4.1-mini	70.45%
Text Persona (Default Temperature)	GPT4.1-mini	71.24%
JSON Persona (Predicted Output)	GPT4.1-mini	69.00%
JSON Persona (Predicted Output)	GPT4.1	71.92%
Random Guessing	—	59.17%

Figure 2: Predictive accuracy



5.2 Aggregate level

We test whether the data simulated from the digital twins replicates the average treatment effects from the 11 classic between-subject studies and the 5 classic within-subject studies included in our experiment. Table 2 shows that for 6 of the 10 results replicated by waves 1-3 and wave 4, the results from the digital twins also replicate the results. For anchoring and adjustment, the digital twins replicate the effect when asking participants to estimate the height of the highest redwood tree. But when asking participants to estimate the number of African countries in the UN, 98.8% of the twins gave the correct answer (54) and no anchoring effect was found. In contrast, only about 10% of humans gave the correct answer (8.79% in wave 2, 10.25% in wave 4). Three other between-subject effects were not replicated. In the outcome bias experiment, participants evaluate a physician's decision to operate on a patient. Humans evaluate the decision more favorably when the operation succeeded than when it failed, despite the risk being greater in the first condition. Overall, about 80% of humans gave a favorable evaluation (78.18% in wave 1, 81.39% in wave 4). In contrast,

digital twins all gave a favorable rating (“correct” to “clearly correct”), with no significant different across conditions. In the sunk cost fallacy experiment, the effect was actually reversed with the digital twins vs. their human counterparts, which we hope future research can explore. In the Allais problem experiment, which tests for violation of the independence axiom of utility theory, all digital twins chose the lower risk - lower reward option over the higher risk - higher reward one. Humans, on the other hand, were much more split in their decisions, and showed systematic differences across conditions (which violates the independence axiom of utility theory). Finally, the base rate fallacy, which was replicated neither in wave 2 nor in wave 4, was not replicated by the digital twins either.

Table 2: Replications of heuristics and biases

Task	Source	Prediction	Waves 1-3	Replicated Wave 4	Twins
<i>Between-subject experiments</i>					
Base rate problem	Kahneman and Tversky (1973)	no difference in prob. assessment when base rate=30 vs. 70	✗	✗	✗
Outcome bias	Baron and Hershey (1988)	average correctness assessment higher in success vs. failure condition	✓	✓	✗
Sunk cost fallacy	Stanovich and West (2008)	average number of purchases higher in sunk cost vs. no sunk cost condition	✓	✓	✗
Allais problem	Stanovich and West (2008)	violation of independence axiom of utility theory (different choices in Form 1 vs. 2)	✓	✓	✗
Framing problem	Tversky and Kahneman (1981)	stronger preference for risky option under loss frame vs. gain frame	✓	✓	✓
Conjunction problem (Linda)	Tversky and Kahneman (1983)	probability assessments higher for feminist bank teller vs. bank teller	✓	✓	✓
Anchoring and adjustment	Tversky and Kahneman (1974) , Epley et al. (2004)	average prediction higher with large vs. small anchor	✓✓	✓✓	✓✗
Absolute vs. relative savings	Stanovich and West (2008)	probability of driving to store higher when discount is larger vs. smaller % of price	✓	✓	✓
Myside bias	Stanovich and West (2008)	average agreement higher for ban of German car in US vs. American car in Germany	✓	✓	✓
Less is More	Stanovich and West (2008)	average attractiveness higher when possibility of loss vs. no possibility of loss	✓	✓	✓
WTA/WTP – Thaler problem	Stanovich and West (2008)	WTA-certainty>WTP-certainty>WTP-noncertainty	✓	✓	✓
<i>Within-subject experiments</i>					
False consensus	Furnas and LaPira (2024)	overpredict (underpredict) public support if own support (oppose)	✓	✓	✓
Nonseparability of risk and benefits judgments	Stanovich and West (2008)	negative correlation between benefits and risks for each item	✓✓✓✗	✓✓✓✗	✓XXX
Omission bias	Stanovich and West (2008)	significant proportion avoid treatment	✓	✓	✗
Probability matching vs. maximizing	Stanovich and West (2008)	significant proportion choose suboptimal strategy	✓	✓	✗
Dominator neglect	Stanovich and West (2008)	significant proportion choose non-normative option	✓	✓	✓

Moving to within-subject experiments, we find that the digital twin results match the human results in two of the five within-subject experiments (see Table 2). In the nonseparability of risk and benefit judgments study, the digital twins judgments display negative correlation as predicted, but the correlation is significant for only one of the items. For probability matching vs. maximizing, the digital twins always selected the normative option while their human counterparts chose the normative option about 30% of the time only. For omission bias, participants were asked whether they would accept a vaccine that prevents catching a flu that has a 10% chance of killing affected patients when the vaccine itself carries a 5% chance of death. While approximately 45% of the human participants (45.10% in wave 2, 44.80% in wave 4) refused the vaccine, only 4.0% of the twins refused the vaccine. This finding echoes our finding related to outcome bias where digital twins were much more favorable to medical professionals compared to their human counterparts.

Finally, we construct average demand curves from the pricing study. Figure A1 shows the average demand curves from the responses from waves 3 vs. wave 4 vs. digital twins. We see that the average demand curves from wave 3 vs. 4 are practically indistinguishable. We find that the average demand curve obtained from the twin is not fully downward sloping, due to the twins' responses to free products. This echoes Gui and Toubia (2023), although digital twins produce demand curves that are downward sloping for positive prices and that are generally closer to the ground truth compared to the demand curves obtained by Gui and Toubia (2023) without such input data.

In sum, while digital twins replicate the majority of between-subject and within-subject effects, there are notable exceptions. Some occur when digital twins fail to mimic the suboptimal or non-normative behaviors of humans, or cannot "unlearn" certain facts (e.g., the number of African countries in the UN). In some areas, twins do match suboptimal human responses (e.g., absolute vs. relative savings, less is more, dominator neglect), raising the broader question of whether digital twins should be seen as "improved" humans or as models that also replicate human deviations from normative behavior and knowledge gaps.

Other deviations appear in the medical domain (e.g., outcome bias, omission bias). Future research should examine whether digital twins are systematically more trusting of the medical profession than humans. Another factor may be that certain topics, such as vaccination, have become highly polarized. This may be considered in light of the result that GPT models tend to struggle to reflect conservative viewpoints (Motoki et al. 2024). For instance, in our false consensus task, while about 45% of humans somewhat or strongly supported increased deportations of those staying in the US illegally (45.42% in wave 1, 45.09% in wave 4), only 25.85% of digital twins did so, and 74.1% strongly or somewhat opposed the measure. More research is needed to systematically study where digital twins diverge from humans, especially in medical and political domains, and to identify other domains where such differences may arise.

6 Conclusion

We present a unique dataset spanning over 500 questions and 2,000 respondents, with high data quality evidenced by sensible correlation patterns, good test-retest accuracy, and replication of known effects. While this resource can broadly benefit business and social science scholars and practitioners, our primary focus is on using it to build digital twins. In initial tests, these twins predict human behavior with out-of-sample accuracy reaching 88% of the test-retest benchmark. Replication of average treatment effects is generally good, though further research is needed to determine if digital twins can capture non-normative behaviors and reflect the full diversity of political and domain-specific views. We also hope that future research will explore the full range of potential applications of digital twins in marketing, business and beyond. Examples include personalization, targeting, product development, professional development and training, advocacy and negotiations, mental health and counseling, etc. The dataset's focus on the US is a potential limitation. Overall, we hope this resource accelerates LLM research as well as business and social science applications while being mindful of societal risks such as dehumanization of research and excessive reliance on AI in decision-making.

Funding and Competing Interests Declarations

All authors certify that they have no affiliations with or involvement in any organization or entity with any financial interest or non-financial interest in the subject matter or materials discussed in this manuscript. The research was supported by one funding institution (kept anonymous for the peer review process).

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Table 3: Appendix Table: Complete list of questions and related measures

Task (source)	#Questions (format)	Extracted measure(s)	Wave(s)
<i>Demographics</i>			
Demographics (Santurkar et al. 2023)	12 (multiple choice)	region, sex, age, education, race, citizenship, marital status, religion, religious attendance, political party, household income, political ideology (categorical)	1
Additional demographics	2 (multiple choice)	household size, employment status (categorical)	1
<i>Personality Traits</i>			
Big 5 personality test (John et al. 1999)	44 (5-point Likert)	extraversion, agreeableness, conscientiousness, neuroticism, openness scores (numerical)	1
Need for cognition scale (Cacioppo et al. 1984)	18 (5-point Likert)	need for cognition score (numerical)	1
Agentic vs. Communal Values scale (Trapnell and Paulhus 2012)	24 (9-point Likert)	agency score, communion score (numerical)	1
Consumer Minimalism scale (Wilson and Bellezza 2022)	12 (5-point Likert)	minimalism score (numerical)	1
Empathy scale (Carré et al. 2013)	20 (5-point Likert)	basic empathy score (numerical)	1
Green values scale (Haws et al. 2014)	6 (5-point Likert)	green score (numerical)	1
Social Desirability scale (Reynolds 1982)	13 (binary choice)	social desirability score (numerical)	2
Conscientiousness scale (Johnson et al. 2019)	8 (9-point Likert)	conscientiousness score (wave 2) (numerical)	2
Anxiety scale (Beck et al. 1988)	21 (4-point Likert)	anxiety score (numerical)	2
Individualism vs. Collectivism scale (Triandis and Gelfand 1998)	16 (5-point Likert)	horizontal/vertical individualism, horizontal/vertical collectivism scores (numerical)	2
Selves questionnaire (Higgins et al. 1985)	3 (open-ended)	n/a	2
Regulatory Focus scale (Fellner et al. 2007)	10 (7-point Likert)	regulatory focus score (numerical)	3
Tightwads vs. Spendthrift scale (Rick et al. 2008)	4 (multiple choice)	tightwads vs. spendthrift score (numerical)	3
Depression scale (Date 1987)	22 (multiple choice)	depression score (numerical)	3
Need for uniqueness scale (Ruvio et al. 2008)	12 (5-point Likert)	need for uniqueness score (numerical)	3
Self-monitoring scale (Lennox and Wolfe 1984)	13 (6-point Likert)	self-monitoring score (numerical)	3
Self-concept clarity scale (Campbell et al. 1996)	12 (5-point Likert)	self-concept clarity score (numerical)	3
Need for closure scale (Roets and Van Hiel 2011)	15 (5-point Likert)	need for closure score (numerical)	3
Maximization scale (Nenkov et al. 2008)	6 (5-point Likert)	maximization score (numerical)	3
<i>Cognitive Abilities</i>			

Cognitive Reflection Test (Krefeld-Schwalb et al. 2024)	4 (open-ended)	CRT score (numerical)	1
Fluid intelligence test (Krefeld-Schwalb et al. 2024)	6 (multiple choice)	fluid intelligence score (numerical)	1
Crystallized intelligence test (Krefeld-Schwalb et al. 2024)	20 (multiple choice)	crystallized intelligence score (numerical)	1
Syllogisms test (Markovits and Nantel 1989)	12 (multiple choice)	syllogism score (numerical)	1
Overconfidence (Dean and Ortoleva 2019)	1 (numerical)	overconfidence score (own predicted-actual score)	1
Overplacement (Dean and Ortoleva 2019)	1 (numerical)	overplacement score (own predicted score-predicted average)	1
Financial literacy test (Johnson et al. 2019)	7 (mult. choice)+1 (num.)	financial literacy score (numerical)	2
Numeracy test (Johnson et al. 2019)	8 (numerical)	numeracy score (numerical)	2
Deductive certainty of Modus Ponens test (Stanovich and West 2008)	4 (binary choice)	deductive certainty score	2
Forward Flow (free associations) (Gray et al. 2019)	20 (open-ended)	forward flow score (average pairwise semantic distance)	2
Wason Selection Task (Klauer et al. 2007)	1 (multiple choice)	Wason Selection Task score (numerical)	3
<i>Economic Preferences</i>			
Ultimatum game (sender) (Güth et al. 1982)	1 (multiple choice)	ultimatum-send (percentage sent)	1
Ultimatum game (receiver) (Güth et al. 1982)	6 (binary choice)	ultimatum-receive (acceptance probability)	1
Mental accounting (Thaler 1985)	4 (binary choice)	mental accounting score (% choices consistent with mental account predictions)	1
Discount (Dean and Ortoleva 2019)	3 (multiple price list)	discount rate (numerical)	2
Present bias (Dean and Ortoleva 2019)	3 (multiple price list)	present bias (numerical)	2
Risk Aversion (Dean and Ortoleva 2019)	3 (uncertainty equivalence)	risk aversion coefficient (numerical)	2
Loss Aversion (Dean and Ortoleva 2019)	4 (uncertainty equivalence)	loss aversion coefficient (numerical)	2
Trust game (sender) (Dean and Ortoleva 2019)	1 (multiple choice)	trust-send (percentage sent)	2
Trust game (receiver) (Dean and Ortoleva 2019)	5 (multiple choice)	Trust-return (average percentage returned)	2
Trust game (sender) thought listing	1 (open-ended)	n/a	2
Trust game (receiver) thought listing	1 (open-ended)	n/a	2
Dictator game (Baron and Hershey 1988)	1 (multiple choice)	dictator-send (percentage sent)	3
Dictator game thought listing	1 (open-ended)	n/a	3
<i>Heuristics and Biases (between subject)</i>			

Base rate problem (Kahneman and Tversky 1973)	1 (slider scale)	average prob. assessment (numerical) in each condition (base rate of 30 vs. 70 engineers)	1, 4
Outcome bias (Baron and Hershey 1988)	1 (7-point Likert)	average correctness assessment (numerical) in each condition (success vs. failure)	1, 4
Sunk cost fallacy (Stanovich and West 2008)	1 (numerical)	average number of purchases (numerical) in each condition (sunk cost yes vs. no)	1, 4
Allais problem (Stanovich and West 2008)	1 (binary choice)	lottery choice probability in each condition (form 1 vs. 2)	1, 4
Framing problem (Tversky and Kahneman 1981)	1 (6-point Likert)	average preference for B vs. A (numerical) in each condition (framing gain vs. loss)	2, 4
Conjunction problem (Linda) (Tversky and Kahneman 1983)	3 (6-point Likert)	average prob. assessment in each condition (feminist bank teller vs. bank teller)	2,4
Anchoring and adjustment (Tversky and Kahneman 1974, Epley et al. 2004)	2 (numerical)	average prediction (numerical) in each condition (with high vs. low anchor)	2, 4
Absolute vs. relative savings (Stanovich and West 2008)	1 (binary choice)	probability of driving to store in each condition (calculator vs. jacket)	2,4
Myside bias (Stanovich and West 2008)	1 (6-point Likert)	average ban agreement (numerical) in each condition (German vs. Ford)	2,4
Less is More (Stanovich and West 2008)	3 (5/6-point Likert)	average attractiveness (numerical) in each condition (Form A vs. B vs. C)	3, 4
WTA/WTP – Thaler problem (Stanovich and West 2008)	1 (multiple choice)	average in each condition (WTP-certainty, WTA-certainty, WTP-noncertainty)	3, 4
<i>Heuristics and Biases (within subject)</i>			
False consensus (Furnas and LaPira 2024)	10 (5-point Likert)+10 (slider)	average predicted public support for each level of own support	1,4
Nonseparability of risk and benefits judgments (Stanovich and West 2008)	8 (7-point Likert)	correlation between benefits and risks for each item	1,4
Omission bias (Stanovich and West 2008)	1 (4-point Likert)	likelihood of taking vaccine (numerical)	2,4
Probability matching vs. maximizing (Stanovich and West 2008)	6-10 (binary choice)	proportion choosing each strategy (Match, Max, other)	3, 4
Dominator neglect (Stanovich and West 2008)	1 (binary choice)	proportion choosing large tray	3,4
<i>Product Preferences</i>			
Pricing study (Gui and Toubia 2023)	40 (binary choice)	demand curve for each product	3,4

414 Appendix

415 A1 Additional tables and figures

Table A1: Demographic characteristics of sample

Category	Count	Percentage
<i>Region</i>		
South	834	40.5%
West	494	24.0%
Midwest	372	18.1%
Northeast	342	16.6%
Pacific	16	0.8%
<i>Sex</i>		
Female	1044	50.7%
Male	1014	49.3%
<i>Age</i>		
18-29	388	18.9%
30-49	735	35.7%
50-64	658	32.0%
65+	277	13.5%
<i>Education</i>		
Less than high school	17	0.8%
High school graduate	272	13.2%
Some college, no degree	468	22.7%
Associate's degree	253	12.3%
College graduate/some postgrad	735	35.7%
Postgraduate	313	15.2%
<i>Race</i>		
White	1361	66.1%
Black	251	12.2%
Hispanic	194	9.4%
Asian	140	6.8%
Other	112	5.4%
<i>Citizenship</i>		
Yes	2054	99.8%
No	4	0.2%
<i>Marital Status</i>		
Married	813	39.5%
Never been married	714	34.7%
Divorced	218	10.6%
Living with a partner	212	10.3%
Widowed	70	3.4%
Separated	31	1.5%
<i>Religion</i>		
Protestant	510	24.8%
Roman Catholic	358	17.4%
Nothing in particular	327	15.9%
Agnostic	311	15.1%
Atheist	216	10.5%
Other	215	10.4%
Jewish	39	1.9%
Buddhist	25	1.2%
Muslim	18	0.9%
Orthodox	17	0.8%
Mormon	15	0.7%
Hindu	7	0.3%
<i>Religious Attendance</i>		
Never	838	40.7%
Seldom	463	22.5%
Once a week	295	14.3%
A few times a year	246	12.0%
Once or twice a month	129	6.3%
More than once a week	87	4.2%
<i>Political Party</i>		
Democrat	847	41.2%
Independent	609	29.6%
Republican	540	26.2%
Something else	62	3.0%
<i>Household Income</i>		
Less than \$30,000	367	17.9%
\$30,000-\$50,000	412	20.0%
\$50,000-\$75,000	411	20.0%
\$75,000-\$100,000	316	15.4%
\$100,000 or more	552	26.8%
<i>Political Ideology</i>		
Moderate	582	28.3%
Liberal	564	27.4%
Conservative	430	20.9%

Category	Count	Percentage
Very liberal	345	16.8%
Very conservative	137	6.7%
<i>Household Size</i>		
1	412	20.0%
2	650	31.6%
3	423	20.6%
4	352	17.1%
More than 4	221	10.7%
<i>Employment Status</i>		
Full-time employment	871	42.3%
Self-employed	280	13.6%
Part-time employment	269	13.1%
Unemployed	249	12.1%
Retired	245	11.9%
Student	78	3.8%
Home-maker	66	3.2%

Table A2: Summary statistics for individual-level measures

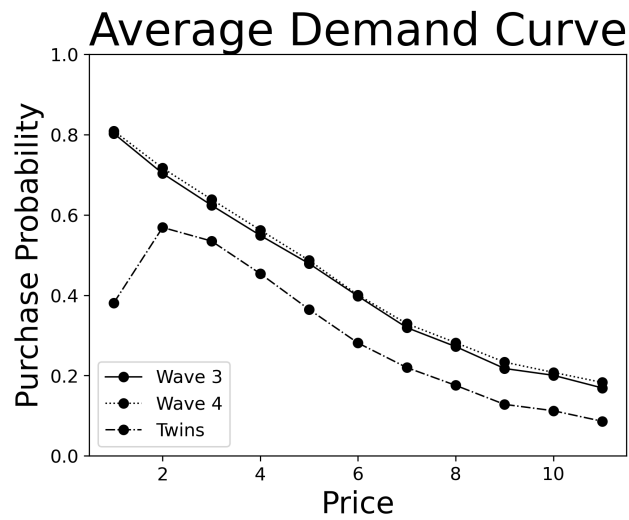
Measure	Average	Std	Median	Min	Max	Theoretical Range
<i>Personality Traits</i>						
extraversion score	2.87	0.93	2.88	1	5	[1,5]
agreeableness score	4.00	0.69	4.00	1.22	5	[1,5]
conscientiousness score	3.93	0.76	4.00	1.11	5	[1,5]
neuroticism score	2.71	1.00	2.63	1	5	[1,5]
openness score	3.75	0.69	3.80	1	5	[1,5]
need for cognition score	3.40	0.83	3.44	1	5	[1,5]
agency score	4.99	1.36	4.83	1	9	[1,9]
communion score	6.94	1.11	7.00	1	9	[1,9]
minimalism score	3.44	0.78	3.50	1	5	[1,5]
basic empathy score	3.88	0.58	3.90	1.50	5	[1,5]
green score	3.51	1.01	3.67	1	5	[1,5]
social desirability score	5.71	3.74	6.00	0	13	[0,13]
conscientiousness score (wave 2)	6.40	2.12	7	0	8	{0,...8}
anxiety score	9.84	9.60	7.00	0	59	{0,...63}
horizontal individualism score	4.23	0.65	4.25	1.25	5	[1,5]
vertical individualism score	2.77	0.90	2.75	1	5	[1,5]
horizontal collectivism score	3.90	0.73	4.00	1	5	[1,5]
vertical collectivism score	3.75	0.85	3.75	1	5	[1,5]
regulatory focus score	4.90	0.64	4.90	2.50	7	[1,7]
tightwad vs. spendthrift score	12.72	4.56	12	4	24	{4,...26}
depression score	11.29	10.25	9	0	55	{0,...61}
need for uniqueness score	2.47	0.89	2.42	1	5	[1,5]
self-monitoring score	2.77	0.46	2.77	0.77	4.38	[0,5]
self-concept clarity score	3.60	0.97	3.75	1	5	[1,5]
need for closure score	3.52	0.65	3.60	1	5	[1,5]
maximization score	3.17	0.65	3.17	1	5	[1,5]
<i>Cognitive Abilities</i>						
CRT score	2.03	1.23	2	0	4	{0,...4}
fluid intelligence score	1.60	1.36	1	0	4	{0,...6}
crystallized intelligence score	6.09	2.38	7	0	9	{0,...20}
syllogism score	6.98	2.22	7	1	11	{0,...12}
overconfidence score	12.18	7.41	13	-18	39	{-42,...42}
overplacement score	1.99	7.93	3.00	-37	40	[-42,42]
financial literacy score	4.99	1.36	5	0	7	{0,...8}
numeracy score	5.43	2.09	6	0	8	{0,...8}
deductive certainty score	3.76	0.64	4	0	4	{0,...4}
forward flow score	0.82	0.05	0.82	0.24	0.93	[0,1]
Wason Selection Task score	2.32	0.68	2	0	4	{0,...4}
<i>Economic Preferences</i>						
ultimatum-send	46.29	20.45	40.00	0	100	[0,100]

ultimatum-receive	80.32	18.04	83.33	0	100	[0,100]
mental accounting score	72.08	24.20	75.00	0	100	[0,100]
discount rate	9.85×10^{13}	4.47×10^{14}	4.83	-1	4.50×10^{15}	$(-1, \infty)$
present bias	0.04	0.11	0	-0.43	0.59	$(-\infty, \infty)$
risk aversion coefficient	0.12	0.24	0.07	-0.67	0.83	$(-\infty, \infty)$
loss aversion coefficient	0.97	0.66	0.89	0.06	6.35	$[0, \infty)$
trust-send	48.47	31.38	40.00	0	100	[0,100]
trust-return	40.67	17.91	45.00	0	100	[0,100]
dictator-send	39.10	18.85	40.00	0	100	[0,100]

Table A3: Examples of significant correlations

Measure 1	Measure 2	Correlation
need for cognition score	openness score	0.62
neuroticism score	depression score	0.62
self-concept clarity score	depression score	-0.55
self-concept clarity score	anxiety score	-0.48
conscientiousness score	depression score	-0.48
agreeableness score	social desirability score	0.44
neuroticism score	social desirability score	-0.41
conscientiousness score	social desirability score	0.37
green score	openness score	0.33
age	self-concept clarity score	0.33

Figure A1: Average demand curves from pricing study



A2 Various approaches of persona construction/simulation

A2.1 Approach details

We experimented with a variety of approaches to construct and simulate LLM personas. Below, we describe each approach corresponding to the rows in Table 1.

- **Text Persona & GPT-4.1-mini:** The full set of survey responses is provided as free-form text, and simulation is performed using GPT-4.1-mini.
- **Text Persona & Gemini-flash2.5:** Identical free-text persona input as above, but simulated with Gemini Flash 2.5 to compare model-dependent behavioral fidelity.
- **JSON Persona & GPT-4.1-mini:** Survey responses are encoded as structured JSON fields instead of text, allowing assessment of the impact of input format on model performance.
- **JSON Persona & GPT-4.1:** Same structured JSON input, but using the full GPT-4.1 model to evaluate the effect of increased model capacity on simulation accuracy.
- **Persona Summary & GPT-4.1-mini:** A concise summary of the persona is provided instead of the complete responses, testing model performance under input length constraints.
- **Persona Summary + JSON Persona & GPT-4.1-mini:** The structured JSON persona is augmented with an appended summary to examine whether a hybrid input format improves results.
- **Text Persona (Reasoning) & GPT-4.1-mini:** The text persona is supplemented with explicit instructions for reasoning before providing answers, following a chain-of-thought prompting approach.
- **Text Persona (Repeating Questions) & GPT-4.1-mini:** The model is prompted to repeat each question and answer choice before responding, ensuring full context is processed during simulation.
- **Text Persona (Default Temperature) & GPT-4.1-mini:** Same textual input as the baseline, but with a default sampling temperature (0.7) to evaluate the impact of increased generation randomness (all other conditions use temperature = 0).
- **JSON Persona (Predicted Output) & GPT-4.1-mini:** Utilizes OpenAI’s “Predicted Output” feature to test whether efficient and accurate structured output can be generated at lower cost.
- **JSON Persona (Predicted Output) & GPT-4.1:** Same as above, but with the full GPT-4.1 model to examine consistency and potential accuracy gains with a larger model.
- **Random Guessing:** A non-informative baseline in which answers are selected uniformly at random, providing a lower-bound reference for accuracy.

System prompt: For all LLM-based simulations, we use the following system prompt:

You are an AI assistant. Your task is to answer the ‘New Survey Question’ as if you are the individual described in the ‘Persona Profile’ (which contains their past survey responses). Remain consistent with the persona’s previous answers and stated characteristics. Carefully follow any instructions provided for the new question, including formatting requirements.

458 A2.2 Text Persona vs JSON Persona vs Persona Summary

459 The **Text Persona** format presents persona information in free-text form, aiming to mimic
460 natural language interaction. An example snippet is shown below:

Text Persona Example

- **Which part of the United States do you currently live in?**

Question Type: Single Choice

Options:

- 1 – Northeast (PA, NY, NJ, RI, CT, MA, VT, NH, ME)
- 2 – Midwest (ND, SD, NE, KS, MN, IA, MO, WI, IL, MI, IN, OH)
- 3 – South (TX, OK, AR, LA, KY, TN, MS, AL, WV, DC, MD, DE, VA, NC, SC, GA, FL)
- 4 – West (WA, OR, ID, MT, WY, CA, NV, UT, CO, AZ, NM)
- 5 – Pacific (HI, AK)

Answer: 3 – South (TX, OK, AR, LA, KY, TN, MS, AL, WV, DC, MD, DE, VA, NC, SC, GA, FL)

- **What is the sex that you were assigned at birth?**

Question Type: Single Choice

Options:

- 1 – Male
- 2 – Female

Answer: 1 – Male

- **How old are you?**

Question Type: Single Choice

Options:

- 1 – 18–29
- 2 – 30–49
- 3 – 50–64
- 4 – 65+

Answer: 1 – 18–29

...

461

462 In contrast, the **JSON Persona** format directly feeds the raw JSON structure of the persona
463 into the LLM. A representative snippet is shown below:

JSON Persona Example

```
[{
  "ElementType": "Block",
  "BlockName": "Demographics",
  "BlockType": "Standard",
  "Questions": [
    {
      "QuestionID": "QID11",
      "QuestionText": "Which part of the United States do you currently live in?",
      "QuestionType": "MC",
      "Options": [
        "Northeast (PA, NY, NJ, RI, CT, MA, VT, NH, ME)",
        "Midwest (ND, SD, NE, KS, MN, IA, MO, WI, IL, MI, IN, OH)",
        "South (TX, OK, AR, LA, KY, TN, MS, AL, WV, DC, MD, DE, VA, NC, SC, GA, FL)",
        "West (WA, OR, ID, MT, WY, CA, NV, UT, CO, AZ, NM)",
        "Pacific (HI, AK)"
      ],
      "Settings": {"Selector": "SAVR", "SubSelector": "TX", "ForceResponse": "ON"},
      "Answers": {
```

464

```

        "SelectedByPosition": 3,
        "SelectedText": "South (TX, OK, AR, LA, KY, TN, MS, AL, WV,
        DC, MD, DE, VA, NC, SC, GA, FL)"
    },
    {
        "QuestionID": "QID12",
        "QuestionText": "What is the sex that you were assigned at birth?",
        "QuestionType": "MC",
        "Options": ["Male", "Female"],
        "Settings": {"Selector": "SAVR", "SubSelector": "TX", "ForceResponse": "ON"},
        "Answers": {"SelectedByPosition": 1, "SelectedText": "Male"}
    },
    {
        "QuestionID": "QID13",
        "QuestionText": "How old are you?",
        "QuestionType": "MC",
        "Options": ["18-29", "30-49", "50-64", "65+"],
        "Settings": {"Selector": "SAVR", "SubSelector": "TX", "ForceResponse": "ON"},
        "Answers": {"SelectedByPosition": 1, "SelectedText": "18-29"}
    }
    ...
  ]}
  ...
]

```

465

466 In addition, we provide a compressed version of the Text Persona—called the *Persona*
 467 *Summary*—by simplifying the questions and summarizing the responses with distributional
 468 information. We envision this format as a more cost-effective option for using personas with
 469 LLMs.

Persona Summary Example

The person's demographics are as follows:

- Geographic region: South (TX, OK, AR, LA, KY, TN, MS, AL, WV, DC, MD, DE, VA, NC, SC, GA, FL)
- Gender: Male
- Age: 18–29
- Education level: Some college, no degree
- Race: White
- Citizen of the US: Yes
- Marital status: Never been married
- Religion: Protestant
- Religious attendance: Once or twice a month
- Political affiliation: Republican
- Income: \$100,000 or more
- Political views: Conservative
- Household size: 4
- Employment status: Student

The person's Big Five personality scores are as follows:

- score_extraversion = 3.5 (75th percentile)
- score_agreeableness = 4.111 (62nd percentile)
- wave1_score_conscientiousness = 4 (53rd percentile)
- score_openness = 3.6 (41st percentile)
- score_neuroticism = 2.5 (47th percentile)
- ...

470

In our evaluations, all formats yield comparable accuracy. We provide all formats in the full dataset to maximize flexibility and usability: <https://huggingface.co/datasets/LLM-Digital-Twin/Twin-2K-500>. The raw JSON format contains the most complete information, with a structured layout that facilitates easy addition, deletion, and retrieval of questions and blocks. The Text Persona format is a human-readable version of the JSON format, designed to resemble natural language prompts compatible with LLM input. The Persona Summary format is a concise alternative to the Text Persona, offering a significantly shorter representation that reduces LLM API usage costs—though at the tradeoff of omitting some detailed information.

A2.3 Discriminative machine learning benchmark and fine tuning

The benchmarks reported above do not use any of the holdout data. Here we examine benchmarks that rely on splitting the holdout data between training and validation samples. These benchmarks represents a fundamentally different task from LLM simulation: while LLM-based digital twins face an out-of-distribution challenge that requires predicting a person’s responses to holdout questions without leveraging any answers from those holdout questions, the benchmarks in this subsection perform an out-of-sample but in-distribution prediction task—they learn from thousands of examples where they observe how other participants’ non-holdout responses relate to their holdout responses. However, these benchmarks are quite useful for evaluating the relevance of the questions we collected and also contextualizing the performance of the digital twin simulation: if these questions were not predictive of individual behaviors even in the presence of actual labeled data, it would be difficult to expect that LLM digital twins could simulate realistic behavior in the absence of such labeled data.

- **XG Boost:** We consider a machine learning benchmark that uses XGBoost to predict responses to holdout questions using numeric non-holdout questions from waves 1-3 through cross-fitting. We implement nested cross-validation where the outer loop uses 5 folds, with each fold serving as a held-out test set once. For each outer training set containing four-fifths of the data, we run an inner 3-fold cross-validation with 30-trial randomized hyperparameter search and early stopping. The best hyperparameters from the inner loop are then used to retrain on the full outer training set before making predictions on the held-out fold. For each of the holdout questions, we train a separate XGBoost model using participants’ responses to all numeric non-holdout questions (demographics, personality scores, cognitive ability scores, economic preference measures) as features. Each model uses approximately 1,600 training examples per fold. XGBoost achieves 75.7% accuracy, compared to 71.72% for the base LLM approach and 81.72% for the test-retest benchmark. This indicates that the non-holdout questions contain predictive information about behavioral responses, reaching 92.6% of the test-retest benchmark. The relatively small gap between XGBoost (75.7%) and LLM digital twins (71.72%) is encouraging, suggesting that digital twins can capture meaningful variation in behavioral responses even without access to holdout question labels.
- **Fine-tuning:** Another approach is to fine-tune the LLM directly on the training data. In our initial trial, we applied supervised fine-tuning (SFT) using OpenAI’s Fine-tuning API with default parameters. The model (GPT-4.1-mini) is lightly fine-tuned on 500 labeled personas. The resulting model achieved 69.61% accuracy, which is actually lower than the original, non-fine-tuned version. This highlights the well-known technical challenges of fine-tuning LLMs. A more rigorous investigation of fine-tuning methods, including both SFT and reinforcement learning, is left for future work.

A3 Detailed instruments and measures

A3.1 Demographic variables

Which part of the United States do you currently live in? [Northeast (PA, NY, NJ, RI, CT, MA, VT, NH, ME); Midwest (ND, SD, NE, KS, MN, IA, MO, WI, IL, MI, IN, OH); South (TX, OK, AR, LA, KY, TN, MS, AL, WV, DC, MD, DE, VA, NC, SC, GA, FL); West (WA, OR, ID, MT, WY, CA, NV, UT, CO, AZ, NM); Pacific (HI, AK)]

What is the sex that you were assigned at birth? [Male; Female]

How old are you? [18-29; 30-49; 50-64; 65+]

What is the highest level of schooling or degree that you have completed? [Less than high school; High school graduate; Some college, no degree; Associate's degree; College graduate/some postgrad; Postgraduate]

What is your race or origin? [White; Black; Asian; Hispanic; Other]

Are you a citizen of the United States? [Yes; No]

Which of these best describes you? [Married; Living with a partner; Divorced; Separated; Widowed; Never been married]

What is your present religion, if any? [Protestant; Roman Catholic; Mormon; Orthodox; Jewish; Muslim; Buddhist; Hindu; Atheist; Agnostic; Other; Nothing in particular]

Aside from weddings and funerals, how often do you attend religious services? [More than once a week; Once a week; Once or twice a month; A few times a year; Seldom; Never]

In politics today, do you consider yourself a [Republican; Democrat; Independent; Something else]

Last year, what was your total family income from all sources, before taxes? [Less than \$30,000; \$30,000-\$50,000; \$50,000-\$75,000; \$75,000-\$100,000; \$100,000 or more]

In general, would you describe your political views as [Very conservative; Conservative; Moderate; Liberal; Very liberal]

Including yourself, how many people currently live in your household? [1; 2; 3; 4; More than 4]

What is your current employment status? [Full-time employment; Part-time employment; Unemployed; Self-employed; Home-maker; Student; Retired]

A3.2 Personality traits

A3.2.1 Big 5 Personality Test (John et al. 1999)

Here are a number of characteristics that may or may not apply to you. Please indicate next to each statement the extent to which you agree or disagree with that statement. I see myself as someone who...

Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3); Agree a little (4); Agree strongly (5)

Items: Is talkative (1); Tends to find fault with others (2); Does a thorough job (3); Is depressed, blue (4); Is original, comes up with new ideas (5); Is reserved (6); Is helpful and unselfish with others (7); Can be somewhat careless (8); Is relaxed, handles stress well (9); Is curious about many different things (10); Is full of energy (11); Starts quarrels with others (12); Is a reliable worker (13); Can be tense (14); Is ingenious, a deep thinker (15); Generates a lot of enthusiasm (16); Has a forgiving nature (17); Tends to be disorganized (18); Worries a lot (19); Has an active imagination (20); Tends to be quiet (21); Is generally trusting (22); Tends to be lazy (23); Is emotionally stable, not easily upset (24); Is inventive (25); Has an assertive personality (26); Can be cold and aloof (27); Perseveres until the task is finished (28); Can be moody (29); Values artistic, aesthetic experiences (30); Is sometimes shy, inhibited (31); Is considerate and kind to almost everyone (32); Does things efficiently (33); Remains calm in tense situations (34); Prefers work that is routine (35); Is outgoing, sociable (36); Is sometimes rude to others (37); Makes plans and follows through with them (38); Gets nervous easily (39); Likes to reflect, play with ideas (40); Has few artistic interests (41); Likes to cooperate with others (42); Is easily distracted (43); Is sophisticated in art, music, or literature (44).

Scores: extraversion: 1, 6R, 11, 16, 21R, 26, 31R, 36; agreeableness: 2R, 7, 12R, 17, 22, 27R, 32,

573 37R, 42; conscientiousness: 3, 8R, 13, 18R, 23R, 28, 33, 38, 43R; openness: 5, 10, 15, 20, 25, 30,
574 35R, 40, 41R, 44; neuroticism: 4, 9R, 14, 19, 24R, 29, 34R, 39.

575 A3.2.2 *Need for cognition scale (Cacioppo et al. 1984)*

576
577 *Here are a number of characteristics that may or may not apply to you. Please indicate next to each*
578 *statement the extent to which you agree or disagree with that statement.*
579 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
580 Agree a little (4); Agree strongly (5)
581 Items: I would prefer complex to simple problems; I like to have the responsibility of
582 handling a situation that requires a lot of thinking; Thinking is not my idea of fun *; I would
583 rather do something that requires little thought than something that is sure to challenge my
584 thinking abilities *; I try to anticipate and avoid situations where there is likely chance I will
585 have to think in depth about something *; I find satisfaction in deliberating hard and for
586 long hours ; I only think as hard as I have to *; I prefer to think about small, daily projects to
587 long-term ones *; I like tasks that require little thought once I've learned them *; The idea
588 of relying on thought to make my way to the top appeals to me; I really enjoy a task that
589 involves coming up with new solutions to problems; Learning new ways to think doesn't
590 excite me very much *; I prefer my life to be filled with puzzles that I must solve; The notion
591 of thinking abstractly appeals to me; I would prefer a task that is intellectual, difficult, and
592 important to one that is somewhat important but does not require too much thought; I feel
593 relief rather than satisfaction after completing a task that requires a lot of mental effort;
594 It's enough for me that something gets the job done; I don't care how or why it works *; I
595 usually end up deliberating about issues even when they do not affect me personally.
596 Score: need for cognition. *: reverse-coded items.

597 A3.2.3 *Agentic vs. Communal Values scale (Trapnell and Paulhus 2012)*

598
599 *Below are 24 different values that people rate of different importance in their lives. FIRST READ*
600 *THROUGH THE LIST to familiarize yourself with all the values. While reading over the list,*
601 *consider which ones tend to be most important to you and which tend to be least important to*
602 *you. After familiarizing yourself with the list, rate the relative importance of each value to you as*
603 *"A GUIDING PRINCIPLE IN MY LIFE." It is important to spread your ratings out as best you*
604 *can—be sure to use some numbers in the lower range, some in the middle range, and some in the*
605 *higher range. Avoid using too many similar numbers. Work fairly quickly.*
606 Response scale: Not Important to me 1: 1; 2; 3; 4; Quite Important to me 5: 5; 6; 7; 8; Highly
607 Important to me 9: 9
608 Items: WEALTH (financially successful, prosperous) (1); PLEASURE (having one's fill
609 of life's pleasures and enjoyments) (2); FORGIVENESS (pardoning others' faults, being
610 merciful) (3); INFLUENCE (having impact, influencing people and events) (4); TRUST
611 (being true to one's word, assuming good in others) (5); COMPETENCE (displaying mastery,
612 being capable, effective) (6); HUMILITY (appreciating others, being modest about oneself)
613 (7); ACHIEVEMENT (reaching lofty goals) (8); ALTRUISM (helping others in need) (9);
614 AMBITION (high aspirations, seizing opportunities) (10); LOYALTY (being faithful to
615 friends, family, and group) (11); POLITENESS (courtesy, good manners) (12); POWER
616 (control over others, dominance) (13); HARMONY (good relations, balance, wholeness)
617 (14); EXCITEMENT (seeking adventure, risk, an exciting lifestyle) (15); HONESTY (being
618 genuine, sincere) (16); COMPASSION (caring for others, displaying kindness) (17); STATUS
619 (high rank, wide respect) (18); CIVILITY (being considerate and respectful toward others)
620 (19); AUTONOMY (independent, free of others' control) (20); EQUALITY (human rights
621 and equal opportunity for all) (21); RECOGNITION (becoming notable, famous, or admired)
622 (22); TRADITION (showing respect for family and cultural values) (23); SUPERIORITY
623 (defeating the competition, standing on top) (24).
624 Scores: agency: 1, 2, 4, 6, 8, 10, 13, 15, 18, 20, 22, 24; communion: 3, 5, 7, 9, 11, 12, 14, 16, 17,
625 19, 21, 23.

626 **A3.2.4 Consumer Minimalism scale (Wilson and Bellezza 2022)**

627

Please indicate your agreement with each of the following statements about yourself.

628 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
629 Agree a little (4); Agree strongly (5)

630 Items: I avoid accumulating lots of stuff (1); I restrict the number of things I own (2); “Less
631 is more” when it comes to owning things (3); I actively avoid acquiring excess possessions
632 (4); I am drawn to visually sparse environments (5); I prefer simplicity in design (6); I keep
633 the aesthetic in my home very sparse (7); I prefer leaving spaces visually empty over filling
634 them (8); I am mindful of what I own (9); The selection of things I own has been carefully
635 curated (10); It is important to me to be thoughtful about what I choose to own (11); My
636 belongings are mindfully selected (12).

637 Score: consumer minimalism (average of all items).
638

639 **A3.2.5 Empathy scale (Carré et al. 2013)**

640

Please indicate your agreement with each of the following statements about yourself.

641 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
642 Agree a little (4); Agree strongly (5)

643 Items: My friends’ emotions don’t affect me much *; After being with a friend who is sad
644 about something, I usually feel sad.; I can understand my friend’s happiness when she/he
645 does well at something.; I get frightened when I watch characters in a good scary movie.; I
646 get caught up in other people’s feelings easily.; I find it hard to know when my friends are
647 frightened. *; I don’t become sad when I see other people crying. *; Other people’s feelings
648 don’t bother me at all. *; When someone is feeling down I can usually understand how
649 they feel.; I can usually work out when my friends are scared.; I often become sad when
650 watching sad things on TV or in films.; I can often understand how people are feeling even
651 before they tell me.; Seeing a person who has been angered has no effect on my feelings. *; I
652 can usually work out when people are cheerful.; I tend to feel scared when I am with friends
653 who are afraid.; I can usually realize quickly when a friend is angry.; I often get swept up in
654 my friends’ feelings.; My friend’s unhappiness doesn’t make me feel anything. *; I am not
655 usually aware of my friends’ feelings. *; I have trouble figuring out when my friends are
656 happy. *
657

658 Score: basic empathy. *: reverse-coded items.

659 **A3.2.6 Green values scale (Haws et al. 2014)**

660

*Here are a number of characteristics that may or may not apply to you. Please indicate next to each
662 statement the extent to which you agree or disagree with that statement.*

663 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
664 Agree a little (4); Agree strongly (5)

665 Items: It is important to me that the products I use do not harm the environment.; I consider
666 the potential environmental impact of my actions when making many of my decisions.;
667 My purchase habits are affected by my concern for our environment.; I am concerned
668 about wasting the resources of our planet.; I would describe myself as environmentally
669 responsible.; I am willing to be inconvenienced in order to take actions that are more
670 environmentally friendly.

671 Score: green (average of all items).

672 **A3.2.7 Social Desirability scale (Reynolds 1982)**

673

*Listed below are a number of statements concerning personal attributes and traits. Read each item
675 and decide whether the statement is true or false as it pertains to your personally.*

676 Response scale: FALSE, TRUE

677 Items: It is sometimes hard for me to go on with my work if I am not encouraged (1); I

678 sometimes feel resentful when I don't get my way (2); On a few occasions, I have given up
679 doing something because I thought too little of my ability (3); There have been times when I
680 felt like rebelling against people in authority even though I knew they were right (4); No
681 matter who I'm talking to, I'm always a good listener (5); There have been occasions when I
682 took advantage of someone (6); I'm always willing to admit when I make a mistake (7); I
683 sometimes try to get even rather than forgive and forget (8); I am always courteous, even
684 to people who are disagreeable (9); I have never been irked when people expressed ideas
685 different from my own (10); There have been times when I was quite jealous of the good
686 fortune of others (11); I am sometimes irritated by people who ask favors of me (12); I have
687 never deliberately said something that hurt someone's feelings (13).
688 Score: social desirability (sum of TRUE responses to items 5, 7, 9, 10, 13 and FALSE responses
689 to items 1, 2, 3, 4, 6, 8, 11, 12).

690 **A3.2.8 Conscientiousness scale (Johnson et al. 2019)**

691
692 *Following are a number of characteristics that may or may not apply to you. Please indicate next to*
693 *each statement the extent to which that statement accurately or inaccurately describes you.*
694 Response scale: Extremely inaccurate: 1; 2; 3; 4; Neither inaccurate nor accurate: 5; 6; 7; 8;
695 Extremely accurate: 9
696 Items: Organized (1); Efficient (2); Systematic (3); Practical (4); Disorganized (5); Sloppy (6);
697 Inefficient (7); Careless (8)
698 Score: conscientiousness scale (wave 2). number of items 1-4 for which response > 5 plus
699 items 5-8 for which response < 5.

700 **A3.2.9 Anxiety scale (Beck et al. 1988)**

701
702 *How much have you been bothered by each of the following symptoms over the past week?*
703 Response scale: Not at all: 0; 1; 2; Severely - I barely could stand it: 3
704 Items: Numbness or tingling; Feeling hot; Wobbliness in legs; Unable to relax; Fear of the
705 worst happening; Dizzy or lightheaded; Unsteady; Terrified; Nervous; Feeling of choking;
706 Hands trembling; Shaky; Fear of losing control; Difficulty breathing; Fear of dying; Scared;
707 Indigestion or discomfort in abdomen; Faint; Face flushed; Sweating (not due to heat); Heart
708 pounding or racing.
709 Score: anxiety score (add up numerical values across items).

710 **A3.2.10 Individualism vs. Collectivism scale (Triandis and Gelfand 1998)**

711
712 *Following are a number of characteristics that may or may not apply to you. Please indicate next to*
713 *each statement the extent to which you agree or disagree with that statement.*
714 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
715 Agree a little (4); Agree strongly (5)
716 Items: I'd rather depend on myself than others (1); I rely on myself most of the time, I rarely
717 rely on others (2); I often do my own thing (3); My personal identity, independent of others,
718 is very important to me (4); It is important for me to do my job better than the others (5);
719 Winning is everything (6); Competition is the law of nature (7); When another person does
720 better than I do, I get tense and aroused (8); If a co-worker gets a prize, I would feel proud
721 (9); The well-being of my coworkers is important to me (10); To me, pleasure is spending
722 time with others (11); I feel good when I cooperate with others (12); Parents and children
723 must stay together as much as possible (13); It is my duty to take care of my family, even
724 when I have to sacrifice what I want (14); Family members should stick together, no matter
725 what sacrifices are required (15); It is important to me that I respect the decision made by
726 my groups (16).
727 Scores: horizontal individualism (items 1-4), vertical individualism (items 5-8), horizontal
728 collectivism (items 9-12), vertical collectivism (items 13-16).

729 **A3.2.11 Selves questionnaire (Higgins et al. 1985)**

730

731 *In this task we would like you to write about the type of person you aspire to be vs. the person you*
732 *ought to be vs. the person you actually are.*

733 *1. Please describe the type of person you aspire to be. That is, write about the traits and behaviors*
734 *you would like ideally to possess, your ultimate goals for yourself. Please write at least 3 sentences.*

735 *2. Please describe the type of person you ought to be. That is, write about the traits and behaviors*
736 *attributes that you should or ought to possess, based on your responsibilities and what other people*
737 *expect from you. Please write at least 3 sentences.*

738 *3. Please describe the type of person you actually are. That is, write about the traits and behaviors*
739 *you actually possess. Please write at least 3 sentences.*

740 *Response format: one text box per question.*

741 **A3.2.12 Regulatory Focus scale (Fellner et al. 2007)**

742

743 *Here are a number of characteristics that may or may not apply to you. Please indicate next to each*
744 *statement the extent to which it is true or untrue.*

745 *Response scale: Disagree untrue (1); Not true (2); Probably not true (3); Neither true nor*
746 *untrue (4); Probably true (5); True (6); Definitely true (7)*

747 *Items: I prefer to work without instructions from others; Rules and regulations are helpful*
748 *and necessary for me; For me, it is very important to carry out the obligations placed on me;*
749 *I generally solve problems creatively; I'm not bothered about reviewing or checking things*
750 *really closely; I like to do things in a new way; I always try to make my work as accurate*
751 *and error-free as possible; I like trying out lots of different things, and am often successful*
752 *in doing so; It is important to me that my achievements are recognized and valued by other*
753 *people; I often think about what other people expect of me.*

754 *Scores: regulatory focus score (average numerical values across items).*

755 **A3.2.13 Tightwads vs. Spendthrift scale (Rick et al. 2008)**

756

757 *Question 1: Which of the following best describes your spending habits?*

758 *Response scale: Tightwad (difficulty spending money): 1; 2; 3; 4; 5; About the same or*
759 *neither: 6; 7; 8; 9; 10; Spendthrift (difficulty controlling spending): 11*

760

761 *Question 2a: Some people have trouble limiting their spending: they often spend money -*
762 *for example on clothes, meals, vacations - when they would do better not to.*

763 *How well does the first description fit you? That is, do you have trouble limiting your spending?*

764 *Response scale: Never (1); Rarely (2); Sometimes (3); Often (4); Always (5)*

765

766 *Question 2b: Other people have trouble spending money. Perhaps because spending money makes*
767 *them anxious, they often don't spend money on things they should spend it on.*

768 *How well does the second description fit you? That is, do you have trouble spending money?*

769 *Response scale: Never (1); Rarely (2); Sometimes (3); Often (4); Always (5)*

770

771 *Question 3: Following is a scenario describing the behavior of two shoppers. After read-*
772 *ing about each shopper, please answer the question that follows.*

773 *Mr. A is accompanying a good friend who is on a shopping spree at a local mall. When they enter a*
774 *large department store, Mr. A sees that the store has a "one-day-only-sale" where everything is*
775 *priced 10-60% off. He realizes that he doesn't need anything, yet can't resist and ends up spending*
776 *almost \$100 on stuff.*

777 *Mr. B is accompanying a good friend who is on a shopping spree at a local mall. When they enter*
778 *a large department store, Mr. B sees that the store has a "one-day-only-sale" where everything*
779 *is priced 10-60% off. He figures that he can get great deals on many items that he needs, yet the*
780 *thought of spending the money keeps him from buying the stuff.*

781 *In terms of your own behavior, who are you more similar to, Mr. A or Mr. B?*

782 *Response scale: Mr. A: 1; 2; About the same or neither: 3; 4; Mr. B: 5*

783
784 Scores: tightwads vs. spendthrift score (add up numerical values across items,
785 questions 2b and 3 are reverse coded).

786 **A3.2.14 Depression scale (Date 1987)**

787
788 *This page contains groups of statements. After reading each group of statements carefully, choose the*
789 *one statement which best describes the way you have been feeling in the past week, including today.*
790 *If several statements within a group seem to apply equally well, select each one. Be sure to read all*
791 *the statements in each group before making your choice.*

792 Items:

- 793 1. I do not feel sad (0); I feel sad (1); I am sad all the time and I can't snap out of it (2); I am
794 so sad or unhappy that I can't stand it (3)
- 795 2. I am not particularly discouraged about the future (0); I feel discouraged about the future
796 (1); I feel that I have nothing to look forward to (2); I feel that the future is hopeless and that
797 things cannot improve (3)
- 798 3. I do not feel like a failure (0); I feel that I have failed more than the average person (1); As
799 I look back on my life, all I can see is a lot of failures (2); I feel I am a complete failure as a
800 person (3)
- 801 4. I get as much satisfaction out of things as I used to (0); I don't enjoy things the way I used
802 to (1); I don't get real satisfaction out of anything anymore (2); I am dissatisfied or bored
803 with everything (3)
- 804 5. I don't feel particularly guilty (0); I feel guilty a good part of the time (1); I feel guilty
805 most of the time (2); I feel guilty all the time (3)
- 806 6. I don't feel I am being punished (0); I feel I may be punished (1); I expect to be punished
807 (2); I feel I am being punished (3)
- 808 7. I don't feel disappointed in myself (0); I feel disappointed in myself (1); I am disgusted
809 with myself (2); I hate myself (3)
- 810 8. I don't feel I am worse than anybody else (0); I am critical of myself for my weaknesses or
811 mistakes (1); I blame myself all the time for my faults (2); I blame myself for everything bad
812 that happens (3)⁷
- 813 10. I don't cry any more than usual (0); I cry more than I used to (1); I cry all the time now
814 (2); I used to be able to cry, but now I can't cry even though I want to (3)
- 815 11. I am no more irritated now than I ever am (0); I get annoyed or irritated more easily than
816 I used to (1); I feel irritated all the time now (2); I don't get irritated at all by the things that
817 used to irritate me (3)
- 818 12. I have not lost interest in other people (0); I am less interested in other people than I used
819 to be (1); I have lost most of my interest in other people (2); I have lost all of my interest in
820 other people (3)
- 821 13. I make decisions about as well as I ever could (0); I put off making decisions more
822 than I used to (1); I have greater difficulty in making decisions than before (2); I can't make
823 decisions at all anymore (3)
- 824 14. I do not feel that I am worthless (0); I don't consider myself as worthwhile and useful as
825 I used to (1); I feel more worthless as compared to other people (2); I feel utterly worthless
826 (3)
- 827 15. I can work about as well as before (0); It takes an extra effort to get started at doing
828 something (1); I have to push myself very hard to do anything (2); I can't do any work at all
829 (2)
- 830 16. I can sleep as well as usual (0); I don't sleep as well as usual (1); I wake up 1-2 hours
831 earlier than usual and find it hard to get back to sleep (2); I wake up several hours earlier
832 than I used to and cannot get back to sleep (3)
- 833 17. I don't get more tired than usual (0); I get tired more easily than I used to (1); I get tired
834 from doing almost anything (2); I am too tired to do anything (3)
- 835 18. My appetite is no worse than usual (0); My appetite is not as good as it used to be (1);
836 My appetite is much worse now (2); I have no appetite at all anymore (3)
- 837 19. I haven't lost much weight, if any, lately (0); I have lost more than 5 pounds (1); I have

⁷Item 9, which is about suicidal thoughts, was skipped.

838 lost more than 10 pounds (2); I have lost more than 15 pounds (3)
839 20. I am no more worried about my health than usual (0); I am worried about physical
840 problems such as aches and pains; or upset stomach; or constipation (1); I am very worried
841 about physical problems and it's hard to think of much else (2); I am so worried about my
842 physical problems that I cannot think about anything else
843 21. I have not noticed any recent change in my interest in sex (0); I am less interested in
844 sex than I used to be (1); I am much less interested in sex now (2); I have lost interest in sex
845 completely (3)
846 22. I am purposely trying to lose weight by eating less. Yes (1); No (0)
847 Score: depression score (add up numerical values across items).

848 **A3.2.15 Need for uniqueness scale (Ruvio et al. 2008)**

849
850 *Here are a number of characteristics that may or may not apply to you. Please indicate next to each*
851 *statement the extent to which you agree or disagree with that statement.*
852 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
853 Agree a little (4); Agree strongly (5)
854 Items: I often combine possessions in such a way that I create a personal image that cannot
855 be duplicated; I often try to find a more interesting version of run-of-the-mill products
856 because I enjoy being original; I actively seek to develop my personal uniqueness by buying
857 special products or brands; Having an eye for products that are interesting and unusual
858 assists me in establishing a distinctive image; When it comes to the products I buy and
859 the situations in which I use them, I have broken customs and rules; I have often violated
860 the understood rules of my social group regarding what to buy or own; I have often gone
861 against the understood rules of my social group regarding when and how certain products
862 are properly used.; I enjoy challenging the prevailing taste of people I know by buying
863 something they would not seem to accept.; When a product I own becomes popular among
864 the general population, I begin to use it less.; I often try to avoid products or brands that I
865 know are bought by the general population.; As a rule, I dislike products or brands that are
866 customarily bought by everyone.; The more commonplace a product or brand is among the
867 general population, the less interested I am in buying it.
868 Score: need for uniqueness score (average numerical values across items).

869 **A3.2.16 Self-monitoring scale (Lennox and Wolfe 1984)**

870
871 *Here are a number of characteristics that may or may not apply to you. Please indicate next to each*
872 *statement the extent to which that statement is true or false.*
873 Response scale: Certainly, always false (0); Generally false (1); Somewhat false, but with
874 exceptions (2); Somewhat true, but with exceptions (3); Generally true (4); Certainly, always
875 true (5)
876 Items: In social situations, I have the ability to alter my behavior if I feel that something else
877 is called for.; I have the ability to control the way I come across to people, depending on the
878 impression I wish to give them.; When I feel that the image I am portraying isn't working,
879 I can readily change it to something that does.; I have trouble changing my behavior to
880 suit different people and different situations. *; I have found that I can adjust my behavior
881 to meet the requirements of any situation I find myself in.; Even when it might be to my
882 advantage, I have difficulty putting up a good front. *; Once I know what the situation calls
883 for, it's easy for me to regulate my actions accordingly.; I am often able to read people's
884 true emotions correctly through their eyes.; In conversations, I am sensitive to even the
885 slightest change in the facial expression of the person I'm conversing with.; My powers of
886 intuition are quite good when it comes to understanding others' emotions and motives.; I
887 can usually tell when others consider a joke to be in bad taste, even though they may laugh
888 convincingly.; I can usually tell when I've said something inappropriate by reading it in
889 the listener's eyes.; If someone is lying to me, I usually know it at once from that person's
890 manner of expression.
891 Score: self-monitoring score (average numerical values across items). *: reverse-scored
892 items.

893 **A3.2.17 Self-concept clarity scale (Campbell et al. 1996)**

894

895 *Please indicate your agreement with each of the following statements about yourself.*

896 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
897 Agree a little (4); Agree strongly (5)

898 Items: My beliefs about myself often conflict with one another. *; On one day I might have
899 one opinion of myself and on another day I might have a different opinion. *; I spend a lot
900 of time wondering about what kind of person I really am. *; Sometimes I feel that I am not
901 really the person that I appear to be. *; When I think about the kind of person I have been
902 in the past, I'm not sure what I was really like. *; I seldom experience conflict between the
903 different aspects of my personality; Sometimes I think I know other people better than I
904 know myself. *; My beliefs about myself seem to change very frequently. *; If I were asked
905 to describe my personality, my description might end up being different from one day to
906 another day. *; Even if I wanted to, I don't think I would tell someone what I'm really like. *;
907 In general, I have a clear sense of who I am and what I am.; It is often hard for me to make
908 up my mind about things because I don't really know what I want. *
909 Score: self-concept clarity score. *: reverse-coded items.

910 **A3.2.18 Need for closure scale (Roets and Van Hiel 2011)**

911

912 *Please indicate your agreement with each of the following statements about yourself.*

913 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
914 Agree a little (4); Agree strongly (5)

915 Items: I don't like situations that are uncertain.; I dislike questions which could be answered
916 in many different ways.; I find that a well ordered life with regular hours suits my temper-
917 ament.; I feel uncomfortable when I don't understand the reason why an event occurred
918 in my life.; I feel irritated when one person disagrees with what everyone else in a group
919 believes.; I don't like to go into a situation without knowing what I can expect from it.;
920 When I have made a decision, I feel relieved.; When I am confronted with a problem, I'm
921 dying to reach a solution very quickly.; I would quickly become impatient and irritated if
922 I would not find a solution to a problem immediately.; I don't like to be with people who
923 are capable of unexpected actions.; I dislike it when a person's statement could mean many
924 different things.; I find that establishing a consistent routine enables me to enjoy life more.; I
925 enjoy having a clear and structured mode of life.; I do not usually consult many different
926 opinions before forming my own view.; I dislike unpredictable situations.
927 Score: need for closure score (average numerical values across items).

928 **A3.2.19 Maximization scale (Nenkov et al. 2008)**

929

930 *Please indicate your agreement with each of the following statements about yourself.*

931 Response scale: Disagree strongly (1); Disagree a little (2); Neither agree nor disagree (3);
932 Agree a little (4); Agree strongly (5)

933 Items: When I am in the car listening to the radio, I often check other stations to see if
934 something better is playing, even if I am relatively satisfied with what I'm listening to.; No
935 matter how satisfied I am with my job, it's only right for me to be on the lookout for better
936 opportunities.; I often find it difficult to shop for a gift for a friend.; When shopping, I have
937 a hard time finding clothing that I really love.; No matter what I do, I have the highest
938 standards for myself.; I never settle for second best.
939 Score: maximization score (average numerical values across items).

940 **A3.3 Cognitive abilities**

941 **A3.3.1 Cognitive Reflection Test (Krefeld-Schwalb et al. 2024)**

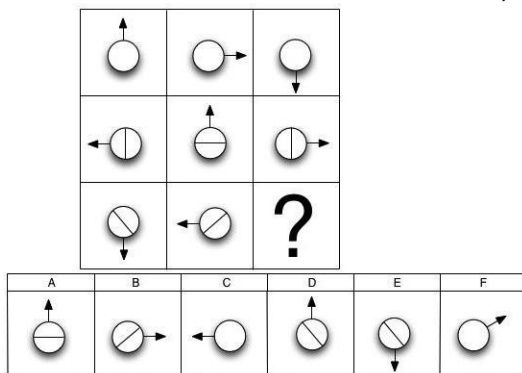
942

943 Questions (response format, correct answer):

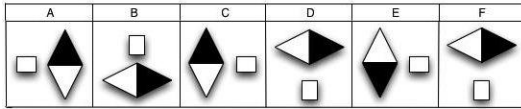
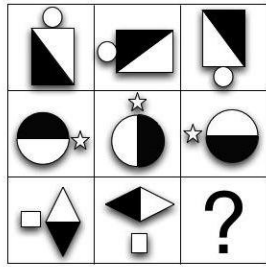
944 Emily's father has three daughters. The first two are named April and May. What is the third
 945 daughter's name? (text, Emily)
 946 How many cubic feet of dirt are there in a hole that is 3' deep x 3' wide x 3' long? Enter a number of
 947 cubic feet. (non-negative number, 0)
 948 If you're running a race and you pass the person in second place, what place are you in? Enter a
 949 number. (non-negative number, 2)
 950 A farmer had 15 sheep and all but 8 died. How many are left? Enter a number (non-negative
 951 number, 8)
 952 Score: CRT (number of correct responses).

953 **A3.3.2 Fluid intelligence test (Krefeld-Schwalb et al. 2024)**

954 Questions (response format, correct answer):
 955 Please indicate which is the best answer to complete the figure below
 956

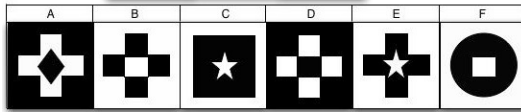
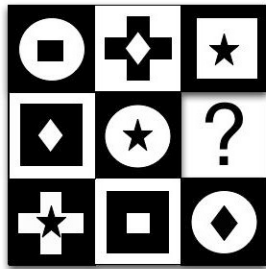


957 (Multiple choice, D)
 958 (When creating digital twins this image was replaced with the following text: Puzzle Grid
 959 (3x3 Matrix). Each cell in a 3x3 matrix contains a circle with an arrow pointing in one of four
 960 directions (up, down, left, right) and sometimes includes a diagonal or horizontal/vertical line inside
 961 the circle. We'll label the grid as follows: Top row (Row 1): Cell 1-1: Plain circle with arrow pointing
 962 up. Cell 1-2: Plain circle with arrow pointing right. Cell 1-3: Plain circle with arrow pointing down.
 963 Middle row (Row 2): Cell 2-1: Circle with a horizontal line, arrow pointing left. Cell 2-2: Circle
 964 with a horizontal line, arrow pointing up. Cell 2-3: Circle with a horizontal line, arrow pointing
 965 right. Bottom row (Row 3): Cell 3-1: Circle with a diagonal line from top-left to bottom-right, arrow
 966 pointing down. Cell 3-2: Circle with a diagonal line from top-left to bottom-right, arrow pointing
 967 left. Cell 3-3: Missing (marked with a question mark – this is what we are trying to determine).
 968 Answer Choices (Labeled A to F): Each option consists of a circle, possibly with an internal line, and
 969 an arrow in a particular direction. A: Circle with horizontal line, arrow pointing up. B: Circle with
 970 diagonal line (top-left to bottom-right), arrow pointing left. C: Plain circle, arrow pointing left. D:
 971 Circle with diagonal line (top-left to bottom-right), arrow pointing right. E: Circle with diagonal line
 972 (top-left to bottom-right), arrow pointing down. F: Plain circle, arrow pointing up-right (diagonal).
 973 The 5 other images in this section replaced similarly)
 974
 975 Please indicate which is the best answer to complete the figure below
 976



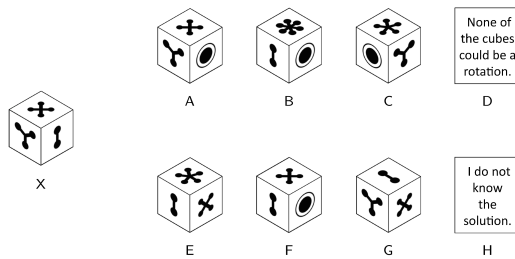
(Multiple choice, C)

Please indicate which is the best answer to complete the figure below



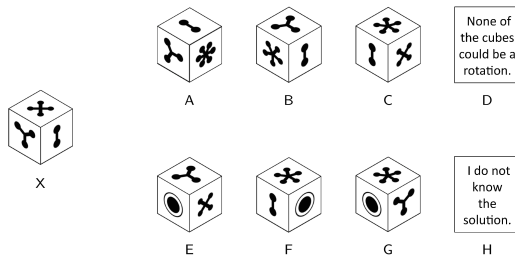
(Multiple choice, B)

All the cubes below have a different image on each side. Select the choice that could represent a rotation of the cube labeled X



(Multiple choice, F)

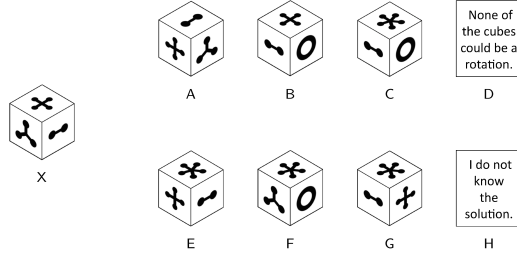
All the cubes below have a different image on each side. Select the choice that could represent a rotation of the cube labeled X



(Multiple choice, A)

All the cubes below have a different image on each side. Select the choice that could represent

995 sent a rotation of the cube labeled X



996

(Multiple choice, B)

998

999 Score: fluid intelligence (number of correct responses).

1000 A3.3.3 Crystallized intelligence test (Krefeld-Schwalb et al. 2024)

1001

1002 Questions (response options with correct answer in *italics*):

1003 *Synonym: Select the word that is most nearly the same in meaning as CONCUR (acquiesce, extricate, divulge, concoct, ransack)*

1005 *Synonym: Select the word that is most nearly the same in meaning as CONFISCATE (harass, repulse, console, appropriate, congregate)*

1007 *Synonym: Select the word that is most nearly the same in meaning as SOLICIT (purge, spurn, entrance, exert, beseech)*

1009 *Synonym: Select the word that is most nearly the same in meaning as FURTIVE (ecstatic, heinous, stealthy, flimsy, facile)*

1011 *Synonym: Select the word that is most nearly the same in meaning as ASTUTE (bizarre, ascetic, sagacious, lineal, irritable)*

1013 *Synonym: Select the word that is most nearly the same in meaning as COVET (crave, claim, avenge, clutch, comply)*

1015 *Synonym: Select the word that is most nearly the same in meaning as OSCILLATE (premeditate, irradiate, vacillate,⁸ recapitulate, furbish)*

1017 *Synonym: Select the word that is most nearly the same in meaning as INDOLENT (contrite, inexhaustible, impervious, arduous, slothful)*

1019 *Synonym: Select the word that is most nearly the same in meaning as DISPARITY (despondency, mediocrity, serenity, incongruity, assiduity)*

1021 *Synonym: Select the word that is most nearly the same in meaning as INDIGENT (refractory, fiscal, destitute, tolerable, diligent)*

1023 *Antonym: Select the word that is most nearly the opposite in meaning to SATIED (famished, finished, finicky, fulfilled, fortunate)*

1025 *Antonym: Select the word that is most nearly the opposite in meaning to COMPLAISANT (distasteful, egoistical, alone, ugly, recalcitrant)*

1027 *Antonym: Select the word that is most nearly the opposite in meaning to ALOOF (happy, dead, gregarious, manly, varied)*

1029 *Antonym: Select the word that is most nearly the opposite in meaning to ABOMINATE (adore, despair, abate, deplore, attach)*

1031 *Antonym: Select the word that is most nearly the opposite in meaning to VERBOSE (garrulous, magnificent, grandiloquent, taciturn, calculating)*

1033 *Antonym: Select the word that is most nearly the opposite in meaning to DEARTH (birth, brevity, abundance, splendor, renaissance)*

1035 *Antonym: Select the word that is most nearly the opposite in meaning to CORPULENT (sallow, affiliated, emaciated, entrepreneur, anemic)*

1037 *Antonym: Select the word that is most nearly the opposite in meaning to GERMANE (teutonic, healthful, irrelevant, massive, puny)*

1039 *Antonym: Select the word that is most nearly the opposite in meaning to VACUOUS (bankrupt, loose, livid, superficial, profound)*

1041 *Antonym: Select the word that is most nearly the opposite in meaning to SPORADIC (germinal,*

⁸Word was misspelled unintentionally in survey.

1042 antiseptic, incessant, summery, wintry)
1043 Score: crystallized intelligence (number of correct responses)

1044 **A3.3.4 Syllogisms test (Markovits and Nantel 1989)**

1045
1046 *Suppose that it is true that:*
1047 *All the XAR's are YOF's.*
1048 *With this in mind, answer the following questions.*
1049
1050 Questions (response options with correct answer in *italics*):
1051 *If a glock is a XAR, you can say: (that it is certain that the glock is a YOF, that it is certain that the*
1052 *glock is not a YOF, that it is not certain whether the glock is a YOF or not)*
1053 *If a glock is not a XAR, you can say: (that it is certain that the glock is a YOF, that it is certain*
1054 *that the glock is not a YOF, that it is not certain whether the glock is a YOF or not)*
1055 *If a koy is a YOF, you can say: (that it is certain that the koy is a XAR, that it is certain that the*
1056 *koy is not a XAR, that it is not certain whether the koy is a XAR or not)*
1057 *If a koy is not a YOF, you can say: (that it is certain that the koy is a XAR, that it is certain that*
1058 *the koy is not a XAR, that it is not certain whether the koy is a XAR or not)*
1059
1060 *You are now going to receive a series of eight problems. You must decide whether the*
1061 *stated conclusion follows logically from the premises or not. You must suppose that the premises are*
1062 *all true and limit yourself only to the information contained in these premises.*
1063
1064 *Premise 1: All things that are smoked are good for the health. Premise 2: Cigarettes are*
1065 *smoked. Conclusion: Cigarettes are good for the health. Does the conclusion follow logically from the*
1066 *premises? (yes, no)*
1067 *Premise 1: All animals love water. Premise 2: Plants do not love water. Conclusion: Plants are not*
1068 *animals. Does the conclusion follow logically from the premises? (yes, no)*
1069 *Premise 1: All animals with four legs are dangerous. Premise 2: Poodles are not dangerous.*
1070 *Conclusion: Poodles do not have four legs. Does the conclusion follow logically from the premises?*
1071 *(yes, no)*
1072 *Premise 1: All eastern countries are communist. Premise 2: China is not an eastern country.*
1073 *Conclusion: China is not communist. Does the conclusion follow logically from the premises? (yes,*
1074 *no)*
1075 *Premise 1: All flowers have petals. Premise 2: Roses have petals. Conclusion: Roses are flowers.*
1076 *Does the conclusion follow logically from the premises? (yes, no)*
1077 *Premise 1: All mammals swim. Premise 2: Whales are mammals. Conclusion: Whales swim. Does*
1078 *the conclusion follow logically from the premises? (yes, no)*
1079 *Premise 1: All unemployed people are poor. Premise 2: Rockefeller is not unemployed. Conclusion:*
1080 *Rockefeller is not poor. Does the conclusion follow logically from the premises? (yes, no)*
1081 *Premise 1: All things that have a motor need oil. Premise 2: Bicycles need oil. Conclusion: Bicycles*
1082 *have motors. Does the conclusion follow logically from the premises? (yes, no)*
1083
1084 Score: syllogism (number of correct responses).

1085 **A3.3.5 Overconfidence (Dean and Ortoleva 2019)**

1086
1087 *Question: You just answered 42 questions that measured your performance on various cognitive*
1088 *tests. How many of these questions do you think you answered correctly? Enter a whole number*
1089 *between 0 and 42.*
1090 *Response format: integer between 0 and 42.*
1091 *Score: overconfidence (belief of own performance on 42 cognitive test questions in wave 1 -*
1092 *actual performance).*

1093 **A3.3.6 Overplacement (Dean and Ortoleva 2019)**

1094

1095 Question: *How many of these questions do you think you people from a representative sample of the*
1096 *US adult population would answer correctly, on average? Enter a whole number between 0 and 42.*
1097 Response format: integer between 0 and 42.

1098 Score: overplacement (belief of own performance on 42 cognitive test questions in wave 1 -
1099 belief of average performance).

1100 **A3.3.7 Financial literacy test (Johnson et al. 2019)**

1101

1102 Multiple-choice questions (response options with correct answer in *italics*):

1103 *Do you think that the following statement is true or false? "A 15-year mortgage typically requires*
1104 *higher monthly payments than a 30-year mortgage, but the total interest paid over the life of the loan*
1105 *will be less."* (True, False)

1106 *Imagine that the interest rate on your saving account was 1% per year and inflation was 2% per*
1107 *year. After 1 year, would you be able to buy: (More than today with the money on this account,*
1108 *Exactly the same as today with the money in this account, Less than today with the money in*
1109 *this account, Do not know)*

1110 *Normally, which asset described below displays the highest fluctuations over time? (Savings account,*
1111 *Stocks, Bonds, Do not know)*

1112 *Do you think that the following statement is true or false? "If you were to invest \$1,000 in a stock*
1113 *mutual fund, it would be possible to have less than \$1,000 when you withdraw your money."* (True,
1114 False)

1115 *When an investor spreads their money among different assets, the risk of losing a lot of money:*
1116 *(Increases, Decreases, Stays the same, Do not know)*

1117 *Considering a long time period (for example 10 or 20 years), which asset described below normally*
1118 *gives the highest return? (Savings account, Stocks, Bonds, Do not know)*

1119 *Do you think that the following statement is true or false? "After age 70 and a half, you have to*
1120 *withdraw at least some money from your 401(k) plan or IRA."* (True, False)⁹

1121 *Suppose you owe \$3,000 on your credit card. You pay a minimum payment of \$30 each month. At*
1122 *an Annual Percentage Rate of 12% (or 1% per month), how many years would it take to eliminate*
1123 *your credit card debt if you made no additional new charges? Enter a number of years, or "never" if*
1124 *the debt will never be eliminated. (open-ended, never).*

1125 Score: financial literacy (number of correct responses)

1126 **A3.3.8 Numeracy test (Johnson et al. 2019)**

1127

1128 Questions (correct answer):

1129 *Imagine that we roll a fair, six-sided die 1,000 times. Out of 1,000 rolls, how many times do you*
1130 *think the die would come up as an even number? Which is the most likely outcome? Enter a number*
1131 *from 0 to 1,000. (500)*

1132 *In the BIG BUCKS LOTTERY, the chance of winning a \$10.00 prize is 1%. What is your best guess*
1133 *about how many people would win a \$10.00 prize if 1,000 people each buy a single ticket from BIG*
1134 *BUCKS? Enter a number from 0 to 1,000. (10)*

1135 *If the chance of getting a disease is 20 out of 100, this would be the same as having a [blank]% chance*
1136 *of getting the disease. Enter a percentage from 0 to 100. (20)*

1137 *In the ACME PUBLISHING SWEEPSTAKES, the chance of winning a car is 1 in 1,000. What*
1138 *percent of tickets of ACME PUBLISHING SWEEPSTAKES win a car? Enter a percentage from 0 to*
1139 *100. (0.1)*

1140 *If the chance of getting a disease is 10%, how many people would be expected to get the disease out of*
1141 *1,000? Enter a number from 0 to 1,000. (100)*

1142 *If it takes 5 machines 5 minutes to make 5 widgets, how long would it take 100 machines to make*
1143 *100 widgets? Enter a number of minutes. (5)*

1144 *A bat and ball cost \$1.10 in total. The bat costs \$1.00 more than the ball. How much does the ball*

⁹Legislation has evolved since the questionnaire was initially developed.

1145 cost? Enter a number in dollars. (0.05)
1146 In a lake, there is a patch of lily pads. Every day, the patch doubles in size. If it takes 48 days for the
1147 patch to cover the entire lake, how long would it take for the patch to cover half of the lake? Enter a
1148 number of days. (47)
1149 Score: numeracy (number of correct responses)

1150 **A3.3.9 Deductive certainty of Modus Ponens test (Stanovich and West 2008)**

1151
1152 You are now going to receive a series of four problems. You must decide whether the stated conclusion
1153 follows logically from the premises or not.
1154 You must suppose that the premises are all true and limit yourself only to the information contained
1155 in these premises.

1156
1157 Questions (correct answer):

1158
1159 Premise 1: If there is a postal strike, then unemployment will double.
1160 Premise 2: There is a postal strike.
1161 Conclusion: Unemployment will double.
1162 Does the conclusion follow logically from the premises? (Yes)

1163
1164 Premise 1: If the winter is harsh, then there will be a flu epidemic.
1165 Premise 2: The winter is harsh.
1166 Conclusion: There will be a flu epidemic.
1167 Does the conclusion follow logically from the premises? (Yes)

1168
1169 Premise 1: If a car is a Honda, then it is expensive.
1170 Premise 2: A car is a Honda.
1171 Conclusion: The car is expensive.
1172 Does the conclusion follow logically from the premises? (Yes)

1173
1174 Premise 1: If a person eats hamburgers, then they will get cancer.
1175 Premise 2: A person eats hamburgers.
1176 Conclusion: The person will get cancer.
1177 Does the conclusion follow logically from the premises? (Yes)

1178
1179 Score: deductive certainty score (number of correct responses)

1180 **A3.3.10 Forward Flow (free associations) (Gray et al. 2019)**

1181 [Each participant randomly assigned to one of the following seed word: candle, snow,
1182 toaster, paper, table, bear]
1183 On this page, starting with the word {seed word}, your job is to write down the next word that
1184 follows in your mind from the previous word. Please put down only single words, and do not use
1185 proper nouns (such as names, brands, etc.). There is no right or wrong answer, just write the words
1186 as they come to your mind.
1187 Response format: 20 text boxes (first pre-populated with the seed word).
1188 Score: forward flow (average pairwise cosine similarity based on the 20 Word2vec embed-
1189 dings).

1190 **A3.3.11 Wason Selection Task (Klauer et al. 2007)**

1191
1192 Imagine you see a number of cards from a set of cards. Each card in the set has a capital letter on one
1193 side and a number on the other. Naturally, only one side is visible in each case.
1194 For the set of cards, a rule has been stated: If there is an A on the letter side of the card, then there is a
1195 3 on the number side.
1196 Four cards were drawn. Below is the information visible for each card (letter or number).
1197 Which of the following card(s) would have to be turned over in order to test the truth or falsity of the

1198 rule?
1199 Options (correct/incorrect): A (correct); F (incorrect); 3 (incorrect); 7 (correct)
1200 Score: Wason Selection Task score (number of correct responses)

1201 **A3.4 Economic preferences**

1202 **A3.4.1 Ultimatum Game (Güth et al. 1982)**

1203
1204 Send question (options): Suppose you were given \$5 and had to offer to another (anonymous)
1205 person a way to split the money. You would propose how much of this money to keep for yourself and
1206 how much to send them.

1207 Then, the other person would have to decide whether or not to accept your offer. If they accept your
1208 offer, you would each receive the amount specified in your offer.

1209 If they reject your offer, you would both receive nothing.

1210 In this scenario, how much would propose to keep for yourself and how much would you propose to
1211 send to the other person? (\$0 for myself, \$5 to the other person; \$1 for myself, \$4 to the other
1212 person.; \$2 for myself, \$3 to the other person; \$3 for myself, \$2 to the other person; \$4 for
1213 myself, \$1 to the other person; \$5 for myself, \$0 to the other person.

1214 Measure: ultimatum-send (% of total amount sent).

1215

1216 Suppose now that you are playing this game as the other person, i.e., the receiver.

1217

1218 For each offer made by the sender, would you accept or reject the offer?

1219 Receive questions (options): If the person offers to keep \$0 for themselves and send me \$5:

1220 (I would accept the offer: \$0 for other, \$5 for me; I would reject the offer: \$0 for both)

1221 If the person offers to keep \$1 for themselves and send me \$4:

1222 (I would accept the offer: \$1 for other, \$4 for me; I would reject the offer: \$0 for both)

1223 If the person offers to keep \$2 for themselves and send me \$3:

1224 (I would accept the offer: \$2 for other, \$3 for me; I would reject the offer: \$0 for both)

1225 If the person offers to keep \$3 for themselves and send me \$2:

1226 (I would accept the offer: \$3 for other, \$2 for me; I would reject the offer: \$0 for both)

1227 If the person offers to keep \$4 for themselves and send me \$1:

1228 (I would accept the offer: \$4 for other, \$1 for me; I would reject the offer: \$0 for both)

1229 If the person offers to keep \$5 for themselves and send me \$0:

1230 (I would accept the offer: \$5 for other, \$0 for me; I would reject the offer: \$0 for both)

1231

1232 Measure: ultimatum-receiver (acceptance probability).

1233 **A3.4.2 Mental accounting (Thaler 1985)**

1234

1235 Questions (options):

1236 Person A was given tickets to lotteries involving the World Series. They won \$50 in one lottery and
1237 \$25 in the other.

1238 Person B was given a ticket to a single, larger World Series lottery. They won \$75.

1239 Who was happier? (Person A, Person B)

1240

1241 Person A received a letter from the IRS saying that they made a minor arithmetical mis-
1242 take on their tax return and owed \$100. They received a similar letter the same day from their state
1243 income tax authority saying they owed \$50. There were no other repercussions from either mistake.

1244 Person B received a letter from the IRS saying that they made a minor arithmetical mistake on their
1245 tax return and owed \$150. There were no other repercussions from either mistake.

1246 Who was more upset? (Person A, Person B)

1247

1248 Person A bought their first New York State lottery ticket and won \$100. Also, in a freak
1249 accident, they damaged the rug in their apartment and had to pay the landlord \$80.

1250 Person B bought their first New York State lottery ticket and won \$20. Who was happier? (Person
1251 A, Person B)

1252
 1253 *Person A's car was damaged in a parking lot. They had to spend \$200 to repair the dam-*
 1254 *age. The same day the car was damaged, they won \$25 in the office football pool.*
 1255 *Person B's car was damaged in a parking lot. They had to spend \$175 to repair the damage.*
 1256 *Who was more upset? (Person A, Person B)*
 1257
 1258 Measure: mental accounting score (percentage of responses consistent with mental
 1259 accounting predictions: A, A, B, B).

1260 **A3.4.3 Discount (Dean and Ortoleva 2019)**

1261
 1262 *Please choose between the following options. For each line in the list, you must choose between the*
 1263 *option on the left and the option on the right. Note that on each line, the option on the left stays the*
 1264 *same while the option on the right gets better as one goes down the list. You can select the option you*
 1265 *would prefer receiving by clicking on the button next to that option.*
 1266
 1267 Choice 1: multiple price list. Left option: \$6.00 in 6 weeks. Right option: \$x in 5
 1268 weeks, with $x \in \{3.00, 4.00, 4.50, 5.00, 5.25, 5.50, 5.75, 6.00, 7.00\}$.
 1269 Measure 1: based on lowest value x for which option on the right is preferred, compute
 1270 equivalent annualized discount rate as $(\frac{6}{x})^{52/1} - 1$. (N\A if no switching).
 1271
 1272 Choice 2: multiple price list. Left option: \$8.00 in 7 weeks. Right option: \$x in 6
 1273 weeks, with $x \in \{4.00, 5.00, 6.00, 7.00, 7.25, 7.50, 7.75, 8.00, 9.00\}$.
 1274 Measure 2: equivalent annualized discount rate: $(\frac{8}{x})^{52/1} - 1$. (N\A if no switching).
 1275
 1276 Choice 3: multiple price list. Left option: \$10.00 in 7 weeks. Right option: \$x in 5
 1277 weeks, with $x \in \{5.00, 6.00, 7.00, 8.00, 8.50, 9.00, 9.25, 9.50, 9.75, 10.00, 11.00\}$.
 1278 Measure 3: equivalent annualized discount rate: $(\frac{10}{x})^{52/2} - 1$. (N\A if no switching).
 1279
 1280 Measure: discount rate (average of three measures).

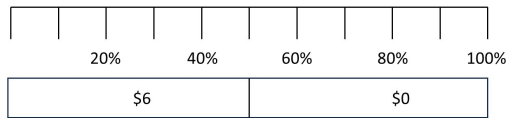
1281 **A3.4.4 Present bias (Dean and Ortoleva 2019)**

1282
 1283 **Three Present Discount questions:**
 1284 *Please choose between the following options. For each line in the list, you must choose between the*
 1285 *option on the left and the option on the right. Note that on each line, the option on the left stays the*
 1286 *same while the option on the right gets better as one goes down the list. You can select the option you*
 1287 *would prefer receiving by clicking on the button next to that option.*
 1288
 1289 Choice 1: multiple price list. Left option: \$6.00 in 1 weeks. Right option: \$x now,
 1290 with $x \in \{3.00, 4.00, 4.50, 5.00, 5.25, 5.50, 5.75, 6.00, 7.00\}$.
 1291 Measure 1: based on lowest value x for which option on the right is preferred, compute
 1292 equivalent annualized discount rate as $(\frac{6}{x})^{52/1} - 1$. (N\A if no switching).
 1293
 1294 Choice 2: multiple price list. Left option: \$8.00 in 1 weeks. Right option: \$x now,
 1295 with $x \in \{4.00, 5.00, 6.00, 7.00, 7.25, 7.50, 7.75, 8.00, 9.00\}$.
 1296 Measure 2: equivalent annualized discount rate: $(\frac{8}{x})^{52/1} - 1$. (N\A if no switching).
 1297
 1298 Choice 3: multiple price list. Left option: \$10.00 in 2 weeks. Right option: \$x now,
 1299 with $x \in \{5.00, 6.00, 7.00, 8.00, 8.50, 9.00, 9.25, 9.50, 9.75, 10.00, 11.00\}$.
 1300 Measure 3: equivalent annualized discount rate: $(\frac{10}{x})^{52/2} - 1$. (N\A if no switching).
 1301
 1302 For each of the three Discount questions, consider the corresponding Present Dis-
 1303 count question. Let x (respectively, z) be the lowest value for which option on the right
 1304 is preferred in the Discount (respectively, Present Discount) question, and let y be the

larger-later amount. Present bias for that pair of question is computed as $\frac{x-z}{y}$. Present bias is measured as the average across the three pairs of questions.

A3.4.5 Risk aversion (Dean and Ortoleva 2019)

Question 1: In this question, the LOTTERY is a 50% chance of winning \$6 and a 50% chance of winning \$0. A graphical representation of the lottery is found below. Suppose you are given the option to exchange this lottery for certain amounts of money. Please choose between the following options. For each line in the list, you must choose between the option on the left and the option on the right. Note that on each line, the option on the left stays the same while the option on the right gets better as one goes down the list.



(When creating digital twins this image was replaced with the following text: The image displays a probability scale from 0% to 100%, marked at intervals of 20%. Below the scale are two side-by-side boxes: one labeled “\$6” on the left that extends from 0% to 50%, and the other labeled “\$3” on the right, that extends from 50% to 100%. Other lottery illustrations were described similarly.)

You can select the option you would prefer receiving by clicking on the button next to that option.

Choice 1: multiple price list. Left option: “Lottery.” Right option: \$x where $x \in \{0.50, 1.00, 1.25, 1.50, 1.75, 2.00, 2.25, 2.50, 2.75, 3.00, 3.25, 3.50, 4.00, 5.00\}$

Measure 1: $\frac{EV-CE}{EV}$ where EV is the lottery’s expected value and CE is the lowest amount for which the option on the right is chosen. (N/A if no switching).

Question 2: same as Question 1, with LOTTERY a 50% chance of winning \$8 and a 50% chance of winning \$2.

Choice 2: Same as Choice 1, where

$x \in \{2.50, 3.00, 3.25, 3.50, 3.75, 4.00, 4.25, 4.50, 4.75, 5.00, 5.25, 5.50, 6.00, 7.00\}$

Measure 2: Same formula as Question 1.

Question 3: same as Question 1, with LOTTERY a 50% chance of winning \$10 and a 50% chance of winning \$0.

Choice 3: Same as Choice 2

Measure 3: Same formula as Question 1.

Measure: risk aversion (average from the three measures).

A3.4.6 Loss aversion (Dean and Ortoleva 2019)

Three questions mirroring the Risk Aversion questions in the loss domain (e.g., first question is: In this question, the LOTTERY is a 50% chance of LOSING \$6 (as indicated by the minus sign before \$6) and a 50% chance of winning \$0. A graphical representation of the lottery is found below. Suppose you are given the option to exchange this lottery for certain amounts of money. The alternative also involves losing money, as indicated by the minus sign. Please choose between the following options. For each line in the list, you must choose between the option on the left and the option on the right. Note that on each line, the option on the left stays the same while the option on the right gets better as one goes down the list. You can select the option you would prefer receiving by clicking on the button next to that option.)

Estimate a constant relative risk aversion (CRRA) coefficient in gain domain based each of the three gain questions.

1354 Estimate a constant relative risk aversion (CRRA) coefficient function in loss do-
 1355 main based on each of the three loss questions.
 1356
 1357 Elicit value x that makes participant indifferent between \$0 for sure and a 50/50
 1358 gamble between x and $-\$8$: *In this question, the LOTTERY is a 50% chance of LOSING \$8*
 1359 *(as indicated by the minus sign before \$8) and a 50% chance of winning a value x . A graphical*
 1360 *representation of the lottery is found below. Suppose you are given the option to exchange this lottery*
 1361 *for the certainty of winning \$0. Please choose between the following options. For each line in the list,*
 1362 *you must choose between the option on the left and the option on the right. Note that on each line,*
 1363 *the option on the left gets better as one goes down the list while the option on the right stays the same.*
 1364 *You can select the option you would prefer receiving by clicking on the button next to that option.*
 1365 Choice: multiple price list. Left option: Lottery with
 1366 $x \in \{7.00, 8.00, 9.00, 10.00, 11.00, 12.00, 13.00, 14.00, 15.00, 16.00, 17.00, 18.00, 19.00, 20.00\}$.
 1367 Right option: \$0.
 1368
 1369 Measure: Loss aversion λ estimated as the additional slope of the utility function
 1370 in the loss domain relative to the gain domain that is necessary to match this last choice,
 1371 conditional on the slopes estimated separately in the two domains. That is, we calculate:
 1372 $\lambda = \frac{-U_G(x)}{U_L(-8)}$ (where x is the lowest amount for which the option on the right is chosen) for
 1373 each of the three pairs of questions, and take the average.

1374 A3.4.7 Trust game (Dean and Ortoleva 2019)

1375
 1376 Questions (response options):
 1377 *Suppose you were given \$5 and had to decide how much of this money to keep for yourself and how*
 1378 *much to send to another (anonymous) person.*
 1379 *Any amount you send to the other person would then be tripled. That is, if you send \$1, this becomes*
 1380 *\$3. If you send \$2, this becomes \$6, etc.*
 1381 *Then, the other person would have to decide how much of that money to keep and how much to return*
 1382 *to you. That is, if you send \$1, this would become \$3 and the other person would have to decide how*
 1383 *much of this \$3 to keep for themselves and how much to send back to you.*
 1384 *In this scenario, how much would keep for yourself and how much would you send to the other*
 1385 *person?*
 1386 *("I would keep \$0 for myself and send \$5 to the other person" to \$5 for myself and \$0 to*
 1387 *other, in \$1 increments.)*
 1388 Measure: trust-send (percentage of total amount sent)
 1389
 1390 *Suppose now that you are playing this game as the other person, i.e., the receiver.*
 1391 *For each amount that you may receive, how much would you keep for yourself and how much would*
 1392 *send back to the other person?*
 1393 *If the person sends me \$5 (which would become \$15):*
 1394 *("I would keep \$0 for myself and send \$15 to the other person" to keep \$15 and send \$0, in*
 1395 *\$1 increments)*
 1396 *If the person sends me \$4 (which would become \$12):*
 1397 *("I would keep \$0 for myself and send \$12 to the other person" to keep \$12 and send \$0, in*
 1398 *\$1 increments)*
 1399 *If the person sends me \$3 (which would become \$9):*
 1400 *("I would keep \$0 for myself and send \$9 to the other person" to keep \$9 and send \$0, in \$1*
 1401 *increments)*
 1402 *If the person sends me \$2 (which would become \$6):*
 1403 *("I would keep \$0 for myself and send \$6 to the other person" to keep \$6 and send \$0, in \$1*
 1404 *increments)*
 1405 *If the person sends me \$1 (which would become \$3):*
 1406 *("I would keep \$0 for myself and send \$3 to the other person" to keep \$3 and send \$0, in \$1*
 1407 *increments)*
 1408
 1409 Measure: trust-receiver (average percentage returned)

1410
1411 Thought listing - sender: *We are now interested in what you were thinking about while*
1412 *deciding how much of the money to keep for yourself and how much to send to another (anonymous)*
1413 *person, and when deciding how much would you keep for yourself and how much would send back to*
1414 *the other person.*
1415 *Any thought is fine; simply list what it was that you were thinking about while answering the*
1416 *questions.*
1417 *Below, please write down the first thought that you had in the first box, the second thought you had*
1418 *in the second box, etc.*
1419 *Please write only one idea per box. You should try to write only the thoughts that you were thinking*
1420 *during the task.*
1421 *Please state your thoughts concisely...one phrase is sufficient. Ignore spelling, grammar, and*
1422 *punctuation.*
1423 *Please be completely honest and list all of thoughts that you had.*
1424 *Don't worry if you don't fill every space. Just write down whatever thoughts you had while making*
1425 *the decision.*
1426 Response format: 6 text boxes (responses optional)
1427
1428 Thought listing - receiver: *Second, please list the thoughts you had when deciding how*
1429 *much would you keep for yourself and how much would send back to the other person:*
1430 Response format: 6 text boxes (responses optional)

1431 **A3.4.8 Dictator Game (Baron and Hershey 1988)**

1432
1433 Question (options): *Suppose you were given \$5 and had to split the money between yourself and*
1434 *another (anonymous) person. You and you only would decide how to split the money, the other*
1435 *person would need to accept your offer.*
1436 *In this scenario, how much would keep for yourself and how much would you send to the other*
1437 *person?*
1438 *(\$0 for myself, \$5 to the other person.; \$1 for myself, \$4 to the other person.; \$2 for myself,*
1439 *\$3 to the other person.; \$4 for myself, \$1 to the other person.; \$5 for myself, \$0 to the other*
1440 *person.)*
1441 Measure: Dictator-send (percentage of total amount sent).
1442
1443 Thought-listing:
1444 *We are now interested in what you were thinking about while deciding how much of the money to*
1445 *keep for yourself and how much to send to another (anonymous) person.*
1446 *Any thought is fine; simply list what it was that you were thinking about while answering the*
1447 *questions.*
1448 *Below, please write down the first thought that you had in the first box, the second thought you had*
1449 *in the second box, etc. Please write only one idea per box. You should try to write only the thoughts*
1450 *that you were thinking during the task.*
1451 *Please state your thoughts concisely...one phrase is sufficient. Ignore spelling, grammar, and*
1452 *punctuation.*
1453 *Please be completely honest and list all of thoughts that you had. Don't worry if you don't fill every*
1454 *space. Just write down whatever thoughts you had while making the decision.*
1455 *Please list the thoughts you had when deciding how much of the money to keep for yourself and how*
1456 *much to send to another (anonymous) person:*
1457 Response format: 6 text boxes (responses optional)

1458 **A3.5 Heuristics and biases - between subject**

1459 **A3.5.1 Base rate problem (Kahneman and Tversky 1973)**

1460
1461 Conditions: 30 Engineers, 70 Engineers
1462 Question: *A panel of psychologist have interviewed and administered personality tests to [30*
1463 *engineers and 70 lawyers, 70 engineers and 30 lawyers], all successful in their respective fields.*

1464 On the basis of this information, thumbnail descriptions of the [30 engineers and 70 lawyers, 70
1465 engineers and 30 lawyers] have been written. Below is one description, chosen at random from the
1466 100 available descriptions.
1467 Jack is a 45-year-old man. He is married and has four children. He is generally conservative, careful,
1468 and ambitious. He shows no interest in political and social issues and spends most of his free time on
1469 his many hobbies which include home carpentry, sailing, and mathematical puzzles.
1470 The probability that Jack is one of the [30, 70] engineers in the sample of 100 is _%. Please indicate
1471 the probability on a scale from 0 to 100.
1472 Response scale: slider (0-100).
1473 Results:
1474 Wave 1: Unlike Tversky and Kahneman (1974) who find no difference between conditions,
1475 we find that the average probability judgment was significantly lower in the “30 Engineers”
1476 condition compared to the “70 Engineers” condition ($Prob_{30} = 52.17\%$, $Prob_{70} = 68.01\%$,
1477 $t = -15.78$, $p < 0.01$).
1478 Wave 4: similar results ($Prob_{30} = 52.39\%$, $Prob_{70} = 70.71\%$, $t = -19.36$, $p < 0.01$).

1479 A3.5.2 Outcome bias (Baron and Hershey 1988)

1480
1481 Conditions: success, failure
1482 Question: A 55-year-old man had a heart condition. He had to stop working because of chest pain.
1483 He enjoyed his work and did not want to stop. His pain also interfered with other things, such
1484 as travel and recreation. A type of bypass operation would relieve his pain and increase his life
1485 expectancy from age 65 to age 70. However, [8% of the people who have this operation die from the
1486 operation itself, 2% of the people who have this operation die from the operation itself]. His physician
1487 decided to go ahead with the operation.
1488 The operation [succeeded, did not succeed and the patient died].
1489 Evaluate the physician’s decision to go ahead with the operation.
1490 Response scale: Incorrect, a very bad decision (-3); Incorrect, all things considered (-2);
1491 Incorrect, but not unreasonable (-1); The decision and its opposite are equally good (0);
1492 Correct, but the opposite would be reasonable tocofo (1); Correct, all things considered (2);
1493 Clearly correct, an excellent decision (3).
1494 Results:
1495 Wave 1: Similar to Baron and Hershey (1988), we find that the average evaluation is more
1496 favorable in the “success” condition compared to the “failure” condition ($M_{success} = 1.66$,
1497 $M_{failure} = 0.88$, $t = 13.55$, $p < 0.001$).
1498 Wave 4: similar results ($M_{success} = 1.64$, $M_{failure} = 1.04$, $t = 11.15$, $p < 0.001$).

1499 A3.5.3 Sunk cost fallacy (Stanovich and West 2008)

1500
1501 Conditions: no, yes
1502 Question - no sunk cost condition: Imagine that Coffee Connection sells coffee for \$1.50 per cup.
1503 Java Coffee, a competitor, sells coffee for just \$2.00 per cup.
1504 Although the Coffee Connection store is ten minutes away by car, Java Coffee is only about 1/2 block
1505 from your apartment.
1506 Assuming that you only buy coffee from these two places and that you like the coffee sold in both
1507 places the same, how many of your next 20 coffee purchases would be from Java Coffee?
1508 Enter a number between 0 and 20.
1509 Question - sunk cost condition: Imagine that you just paid \$50 for a Coffee Connection discount
1510 card that allows you to buy coffee for 50% off the regular price of \$3.00 (i.e., you pay \$1.50).
1511 Soon after you purchased the Coffee Connection discount card, Java Coffee, a competitor, opened a
1512 new store that sells coffee for just \$2.00 per cup.
1513 Although the Coffee Connection store is ten minutes away by car, Java Coffee is only about 1/2 block
1514 from your apartment.
1515 Assuming that you only buy coffee from these two places and that you like the coffee sold in both
1516 places the same, how many of your next 20 coffee purchases would be from Java Coffee?
1517 Enter a number between 0 and 20.
1518 Response format: integer between 0 and 20.

1519
1520 Results:
1521 Wave 1: Similar to Stanovich and West (2008), we find that the average number of
1522 purchases is lower in the “sunk cost” condition compared to the “no sunk cost” condition
1523 ($M_{\text{sunk cost}} = 10.64$, $M_{\text{no sunk cost}} = 14.88$, $t = 16.20$, $p < 0.001$).
1524 Wave 4: similar results ($M_{\text{sunk cost}} = 11.01$, $M_{\text{no sunk cost}} = 14.46$, $t = 14.07$, $p < 0.001$).

1525 **A3.5.4 Allais problem (Stanovich and West 2008)**

1526
1527 Conditions: Form 1, Form2
1528 Choice between two gambles:
1529 Form 1:
1530 One million dollars for sure (A)
1531 89% probability of one million dollars, 10% probability of five million dollars, 1% probability
1532 of nothing (B)
1533
1534 Form 2:
1535 11% probability of one million dollars, 89% probability of nothing (C)
1536 10% probability of five million dollars, 90% probability of nothing (D)
1537
1538 Results:
1539 Wave 1: Similar to Stanovich and West (2008), we find that a significant majority of
1540 participants chose Option A in Form 1 ($\text{Prob}(A) = 69.2\%$, $p < 0.001$), and a significant
1541 majority chose Option D in Form 2 ($\text{Prob}(D) = 57.2\%$, $p < 0.001$).
1542 Wave 4: similar results ($\text{Prob}(A) = 63.6\%$, $p < 0.001$, $\text{Prob}(D) = 62.3\%$, $p < 0.001$).

1543 **A3.5.5 Framing problem (Tversky and Kahneman 1981)**

1544
1545 Conditions: gain framing, loss framing
1546 Question: Imagine that the U.S. is preparing for the outbreak of an unusual disease, which is
1547 expected to kill 600 people.
1548 Two alternative programs to combat the disease have been proposed. Assume that the exact scientific
1549 estimate of the consequences of the programs are as follows: If Program A is adopted, [200 people will
1550 be saved, 400 people will die].
1551 If Program B is adopted, there is 1/3 probability that [600 people will be saved, nobody will die] and
1552 2/3 probability that [no people will be saved, 600 people will die].
1553 Which of the two programs would you favor?
1554 Response scale: I strongly favor program A (1), I favor program A (2), I slightly favor
1555 program A (3), I slightly favor program B (4), I favor program B (5), I strongly favor
1556 program B (6)
1557
1558 Results:
1559 Wave 1: Similar to Tversky and Kahneman (1981), we find that the loss frame resulted in a
1560 greater preference for the risky option B ($M_{\text{gain}} = 2.85$, $M_{\text{loss}} = 3.84$, $t = -17.35$, $p < 0.001$).
1561 Wave 4: Similar results ($M_{\text{gain}} = 2.83$, $M_{\text{loss}} = 3.76$, $t = -17.25$, $p < 0.001$).

1562 **A3.5.6 Conjunction problem (Linda) (Tversky and Kahneman 1983)**

1563
1564 Conditions: bank teller, feminist bank teller
1565 Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student,
1566 she was deeply concerned with issues of discrimination and social justice, and also participated in
1567 anti-nuclear demonstrations.
1568 Please complete the statements below.
1569 Response scale: Extremely improbable (1), Very improbable (2), Somewhat probable (3),
1570 Moderately probable (4), Very probable (5), Extremely probable (6)
1571 Items: It is __ that Linda is a teacher in an elementary school; It is __ that Linda works in a bookstore

1572 and takes Yoga classes; It is ... that Linda is [a bank teller, a bank teller and is active in the feminist
1573 movement]

1574

1575 Results:

1576 Wave 1: consistent with Tversky and Kahneman (1983), we find that Linda was
1577 judged more probably a feminist bank teller than a bank teller ($M_{bank\ teller} = 2.43$,
1578 $M_{feminist\ bank\ teller} = 3.38$, $t = -18.83$, $p < 0.001$)

1579 Wave 2: similar results ($M_{bank\ teller} = 2.52$, $M_{feminist\ bank\ teller} = 3.35$, $t = -17.82$, $p < 0.001$)

1580 A3.5.7 Anchoring and adjustment (Tversky and Kahneman 1974, Epley et al. 2004)

1581

1582 Conditions: large anchor, small anchor

1583

1584 Question 1: Do you think there are more or fewer than [65,12] African countries in the
1585 United Nations? (more, fewer)

1586 Question 2: How many African countries do you think are in the United Nations? (numerical
1587 answer)

1588

1589 Results:

1590 Wave 1: consistent with Tversky and Kahneman (1974), we find that the larger anchor
1591 resulted in higher estimates of the number of African countries in the United Nations
1592 ($M_{large\ anchor} = 48.22$, $M_{small\ anchor} = 26.36$, $t = 4.57$, $p < 0.001$).

1593 Wave 1: similar results ($M_{large\ anchor} = 50.82$, $M_{small\ anchor} = 32.02$, $t = 13.57$, $p < 0.001$).

1594

1595 Question 1: Is the tallest redwood tree in the world more or less than [1000,85] feet tall?
1596 (more, less)

1597 Question 2: How tall do you think the tallest redwood tree in the world is? Enter a number of feet.
1598 (numerical answer)

1599

1600 Results:

1601 Wave 1: consistent with Tversky and Kahneman (1974), we find that the larger anchor
1602 resulted in higher estimates of the height of the tallest redwood tree in the world
1603 ($M_{large\ anchor} = 839.18$, $M_{small\ anchor} = 165.00$, $t = 22.03$, $p < 0.001$).

1604 Wave 2: similar results ($M_{large\ anchor} = 824.01$, $M_{small\ anchor} = 213.17$, $t = 20.80$, $p < 0.001$).

1605 A3.5.8 Absolute vs. relative savings (Stanovich and West 2008)

1606

1607 Conditions: large percentage (calculator), small percentage (jacket)

1608 Question: Imagine that you go to purchase a [calculator for \$30, jacket for \$250].

1609 The [calculator,jacket] salesperson informs you that the [calculator,jacket] you wish to buy is on sale
1610 for [\$20,\$240] at the other branch of the store which is ten minutes away by car.

1611 Would you drive to the other store? (Yes, No)

1612

1613 Results:

1614 Wave 1: consistent with Stanovich and West (2008), we find that more participants were
1615 willing to make the trip to save \$10 for the calculator (large percentage) than for the
1616 jacket (small percentage) ($Prop_{large\ percentage} = 0.74$, $Prop_{small\ percentage} = 0.34$, $\chi^2 = 319.10$,
1617 $p < 0.001$).

1618 Wave 4: similar results ($Prop_{large\ percentage} = 0.73$, $Prop_{small\ percentage} = 0.29$, $\chi^2 = 388.43$,
1619 $p < 0.001$).

1620 A3.5.9 Myside bias (Stanovich and West 2008)

1621

1622 Conditions: German car, Ford Explorer

1623 Question: According to a comprehensive study by the U.S. Department of Transportation, [a

particular German car is, Ford Explorers are] 8 times more likely than a typical family car to kill occupants of another car in a crash.
The [U.S. Department of Transportation, Department of Transportation in Germany] is considering recommending a ban on the sale of [this German car, the Ford Explorer in Germany].
Do you think that [the U.S., Germany] should ban the sale of the [German car, Ford Explorer]?
Response scale: definitely no (1), no (2), probably no (3), probably yes (4), yes (5), definitely yes (6)

Results:

Wave 1: consistent with Stanovich and West (2008), we find that participants were more likely to think that the German car should be banned in the U.S. than they were to think that the Ford Explorer should be banned in Germany ($M_{\text{German car}} = 4.46$, $M_{\text{Ford Explorer}} = 4.11$, $t = 5.86$, $p < 0.001$).

Wave 4: similar results ($M_{\text{German car}} = 4.54$, $M_{\text{Ford Explorer}} = 4.10$, $t = 7.81$, $p < 0.001$).

A3.5.10 Less is More (Stanovich and West 2008)

Conditions: Form A, Form B, Form C.

Question 1: Please rate your level of disagreement or agreement with the following statement:
"I would find a game that had a 7/36 chance of winning \$9 and a 29/36 chance of [winning nothing, losing \$0.05, losing \$0.25] extremely attractive."

Question 2: Imagine that highway safety experts have determined that a substantial number of people are at risk of dying in a type of automobile fire. A requirement that every car have a built-in fire extinguisher (estimated cost, \$300) would save [the 150 people, 98% of the 150 people, 95% of the 150 people] who would otherwise die every year in this type of automobile fire.

Rate the following statement about yourself: I would be supportive of this requirement.

Question 3: You have recently graduated from university, obtained a good job, and are buying a new car. A newly designed seatbelt has just become available that would save the lives of [the 500 drivers, 98% of the 500 drivers, 95% of the 500 drivers] a year who are involved in a type of head-on-collision. (Approximately half of these fatalities involve drivers who were not at fault.) The newly designed seatbelt is not yet standard on most car models. However, it is available as a \$500 option for the car model that you are ordering.

How likely is it that you would order your new car with this optional seatbelt?"

Response scale (common for all three questions): Disagree strongly (1), Disagree a little (2), Neither agree nor disagree (3), Agree a little (4), Agree strongly (5)

Results:

Wave 1: Consistent with Stanovich and West (2008), in each question we find that the option with no possibility of loss (Form A) was rated as less appealing than either of the options that contained the possibility of a loss ($M_A^1 = 2.06$, $M_B^1 = 2.89$, $M_C^1 = 2.86$, $F(2, 2055) = 87.70$, $p < 0.001$; $M_A^2 = 4.03$, $M_B^2 = 4.29$, $M_C^2 = 4.30$, $F(2, 2055) = 13.75$, $p < 0.001$; $M_A^3 = 4.44$, $M_B^3 = 4.75$, $M_C^3 = 4.74$, $F(2, 2055) = 11.19$, $p < 0.001$).

Wave 4: similar results ($M_A^1 = 2.15$, $M_B^1 = 3.15$, $M_C^1 = 3.01$, $F(2, 2055) = 112.01$, $p < 0.001$; $M_A^2 = 3.97$, $M_B^2 = 4.22$, $M_C^2 = 4.27$, $F(2, 2055) = 14.63$, $p < 0.001$; $M_A^3 = 4.44$, $M_B^3 = 4.76$, $M_C^3 = 4.79$, $F(2, 2055) = 14.41$, $p < 0.001$).

A3.5.11 WTA/WTP – Thaler problem (Stanovich and West 2008)

Conditions: WTP-certainty, WTA-certainty, WTP-noncertainty

Question:

WTP-certainty: Imagine that when you went to the movies last week, you were inadvertently exposed to a rare and fatal virus.

The possibility of actually contracting the disease is 1 in 1,000, but once you have the illness there is no known cure.

On the other hand, you can, readily and now, be given an injection that stops the development of the

illness.
 Unfortunately, these injections are only available in very small quantities and are sold to the highest bidder.
 What is the highest price you would be prepared to pay for such an injection? [You can get a long-term, low-interest loan if needed.]:

WTA-certainty: Imagine that a group of research scientists in the School of Medicine are running a laboratory experiment on a vaccine for a rare and fatal virus.
 The possibility of actually contracting the disease from the vaccine is 1 in 1,000, but once you have the disease there is no known cure.
 The scientists are seeking volunteers to test the vaccine on.
 What is the lowest amount that you would have to be paid before you would take part in this experiment?

WTP-noncertainty: Imagine that when you went to the movies last week, you were inadvertently exposed to a rare and fatal virus.
 The possibility of actually contracting the disease is 4 in 1,000, but once you have the illness there is no known cure.
 On the other hand, you can, readily and now, be given an injection that reduces the possibility of contracting the disease to 3 in 1,000.
 Unfortunately, these injections are only available in very small quantities and are sold to the highest bidder.
 What is the highest price you would be prepared to pay for such an injection?

Response scale: \$10 (1), \$100 (2), \$1,000 (3), \$10,000 (4), \$50,000 (5), \$100,000 (6), \$250,000 (7), \$500,000 (8), \$1,000,000 (9), \$5,000,000 or more (10)

Results:
 Wave 1: consistent with Stanovich and West (2008), we find that the mean score in the WTA-certainty condition was significantly higher than the mean score in the WTP-certainty condition ($M_{WTA-certainty} = 6.82$, $M_{WTP-certainty} = 3.27$, $t = 26.25$, $p < 0.001$), and that the mean score in the WTP-certainty condition was significantly higher than the mean score in the WTP-noncertainty condition ($M_{WTP-certainty} = 3.27$, $M_{WTP-noncertainty} = 2.20$, $t = 11.36$, $p < 0.001$).
 Wave 4: similar results ($M_{WTA-certainty} = 7.24$, $M_{WTP-certainty} = 3.23$, $t = 31.15$, $p < 0.001$; $M_{WTP-certainty} = 3.23$, $M_{WTP-noncertainty} = 2.20$, $t = 11.22$, $p < 0.001$).

A3.6 Heuristics and biases - within Subject

A3.6.1 False consensus (Furnas and LaPira 2024)

Self question (asked just before economic preference questions in Wave 1, first question in Wave 4): Would you support or oppose...

Response scale: Strongly oppose, Somewhat oppose, Neither oppose nor support, Somewhat support, Strongly support

Items: Placing a tax on carbon emissions?; Ensuring 40% of all new clean energy infrastructure development spending goes to low-income communities?; Federal investments to ensure a carbon-pollution free electricity sector by 2035?; A 'Medicare for All' system in which all Americans would get healthcare from a government-run plan?; A 'public option', which would allow Americans to buy into a government-run healthcare plan if they choose to do so?; Immigration reforms that would provide a path to U.S. citizenship for undocumented immigrants currently in the United States?; A law that requires companies to provide paid family leave for parents?; A 2% tax on the assets of individuals with a net worth of more than \$50 million?; Increasing deportations for those in the US illegally?; Offering seniors healthcare vouchers to purchase private healthcare plans in place of traditional medicare coverage?

Public question (last question in Wave 1, last question before pricing study in Wave 4): What

percentage of the public do you think supports the following policies? For each policy, choose a number from 0% to 100%.

Response scale: slider (0-100)

Items: same as self question

Analysis: we run a two-way fixed effect regression

$$Y_{ip} = \beta_1 \text{StrongOpp}_{ip} + \beta_2 \text{SomewhatOpp}_{ip} + \beta_3 \text{SomewhatSupp}_{ip} + \beta_4 \text{StrongSupp}_{ip} + \alpha_i + \gamma_p + \epsilon_{ip}$$
, where Y_{ip} is respondent i 's misperception of public support for policy p (predicted-actual public support, with actual support being the proportion of participants who somewhat or strongly support the policy), StrongOpp_{ip} etc. are dummy variables indicating i 's support for p (with "neither oppose nor support" as the reference), α_i is a participant fixed effect, and γ_p is a policy fixed effect.

Results:

Wave 1: consistent with Furnas and LaPira (2024), we find that the more participants support a policy, the more they believe others support it ($\beta_1 = -13.07$, 95% CI=[-14.15, -11.99]; $\beta_2 = -6.38$, 95% CI=[-7.43, -5.36]; $\beta_3 = 8.04$, 95% CI=[7.25, 8.82]; $\beta_4 = 16.65$, 95% CI=[15.84, 17.46]).

Wave 4: similar results ($\beta_1 = -13.62$, 95% CI=[-14.69, -12.55]; $\beta_2 = -5.69$, 95% CI=[-6.68, -4.71]; $\beta_3 = 9.53$, 95% CI=[8.79, 10.28]; $\beta_4 = 18.36$, 95% CI=[17.58, 19.14]).

A3.6.2 Nonseparability of risk and benefits judgments (Stanovich and West 2008)

Benefits: Please rate the following technology or products from "not at all beneficial" to "extremely beneficial"

Response scale: not at all beneficial (1), low benefit (2), slightly beneficial (3), neutral (4), moderately beneficial (5), very beneficial (6), extremely beneficial (7)

Items: bicycles, alcoholic beverages, chemical plants, pesticides

Risks: Please rate the following technology or products from "not at all risky" to "extremely risky"

Response scale: not at all risky (1), low risk (2), slightly risky (3), neutral (4), moderately risky (5), very risky (6), extremely risky (7)

Items: same as benefits question

Results:

Wave 1: we compute the correlation between benefit for each of the four items. Consistent with Stanovich and West (2008), we find significant negative correlations for alcoholic beverages ($r = -0.33$, $t = -15.97$, $p < 0.001$), chemical plants ($r = -0.29$, $t = -13.76$, $p < 0.001$) and pesticides ($r = -0.37$, $t = -18.21$, $p < 0.001$). However, the correlation for bicycles was close to 0 ($r = 0.00$, $t = 0.001$, $p = 1$). Note that Stanovich and West (2008) find that the correlation for bicycles is significant only in their High-SAT group.

Wave 4: similar results ($r_{\text{alcohol}} = -0.36$, $t = -17.31$, $p < 0.001$; $r_{\text{chemical}} = -0.31$, $t = -14.57$, $p < 0.001$; $r_{\text{pesticides}} = -0.37$, $t = -18.26$, $p < 0.001$; $r_{\text{bicycle}} = -0.01$, $t = -0.22$, $p = 0.83$).

A3.6.3 Omission bias (Stanovich and West 2008)

Question: Imagine that there will be a deadly flu going around your area next winter. Your doctor says that you have a 10% chance (10 out of 100) of dying from this flu.

However, a new flu vaccine has been developed and tested. If taken, the vaccine prevents you from catching the deadly flu.

However, there is one serious risk involved with taking this vaccine. The vaccine is made from a somewhat weaker type of flu virus, and there is a 5% (5 out of 100) risk of the vaccine causing you to die from the weaker type of flu.

Imagine that this vaccine is completely covered by health insurance. If you had to decide now, which would you choose?

Response scale: I would definitely not take the vaccine. I would thus accept the 10% chance

1788 of dying from this flu. (1); I would probably not take the vaccine. I would thus accept the
1789 10% chance of dying from this flu. (2); I would probably take the vaccine. I would thus
1790 accept the 5% chance of dying from the weaker flu in the vaccine (3); I would definitely take
1791 the vaccine. I would thus accept the 5% chance of dying from the weaker flu in the vaccine.
1792 (4)

1793 Results:

1795 Wave 2: Consistent with Stanovich and West (2008), we find that a significant proportion of
1796 participants ($Prop_{avoid} = 0.45$, 95% CI=[0.43,0.47]) displayed omission bias, i.e., chose to
1797 avoid the treatment (answer 1 or 2 vs. 3 or 4).

1798 Wave 4: similar results ($Prop_{avoid} = 0.45$, 95% CI=[0.43,0.47]).

1799 A3.6.4 Probability matching vs. maximizing (Stanovich and West 2008)

1800

1801 Conditions: card problem, dice problem

1802

1803 Card problem: Consider the following hypothetical situation:

1804 textitA deck with 10 cards is randomly shuffled 10 separate times. The 10 cards are
1805 composed of 7 cards with the number “1” on the down side and 3 cards with the number
1806 “2” on the down side.

1807 Each time the 10 cards are reshuffled, your task is to predict the number on the down side of the top
1808 card.

1809 Imagine that you will receive \$100 for each downside number you correctly predict, and that you
1810 want to earn as much money as possible.

1811 What would you predict after ...

1812 10 items: shuffle #1,...shuffle #10

1813 Response scale (choose one): 1, 2

1814 Measure: participant classified as using the MAX strategy (normative) if chose 1 ten times,
1815 the MATCH strategy if chose 1 seven times and 2 three times, and the OTHER strategy if
1816 made any other set of choices.

1817

1818 Dice problem: Consider the following hypothetical situation:

1819 Consider the following situation:

1820 A die with 4 red faces and 2 green faces will be rolled 6 times.

1821 Before each roll you will be asked to predict which color (red or green) will show up once the die is
1822 rolled.

1823 Which color is most likely to show up after ...

1824 6 items: roll #1,...roll #6

1825 Response scale (choose one): red, green

1826 Measure: participant classified as using the MAX strategy (normative) if chose red six times,
1827 the MATCH strategy if chose red four times and green two times, and the OTHER strategy
1828 if made any other set of choices.

1829

1830 Results:

1831 Wave 1: consistent with Stanovich and West (2008), we find that in both conditions, a
1832 significant proportion of participants chose a non-normative strategy ($P_{MAX}^{card} = 0.36$, 95%
1833 CI=[0.33,0.39]; $P_{MAX}^{dice} = 0.30$, 95% CI=[0.27,0.33]).

1834 Wave 4: similar results ($P_{MAX}^{card} = 0.36$, 95% CI=[0.34,0.39]; $P_{MAX}^{dice} = 0.29$, 95% CI=[0.27,0.32]).

1835 A3.6.5 Dominator neglect (Stanovich and West 2008)

1836

1837 Question: Assume that you are presented with two trays of black and white marbles, a large tray
1838 that contains 100 marbles and a small tray that contains 10 marbles. The marbles are spread in a
1839 single layer in each tray.

1840 You must draw out one marble (without peeking, of course) from either tray. If you draw a black
1841 marble you win \$2. Consider a condition in which the small tray contains 1 black marble and 9

1842 white marbles, and the large tray contains 8 black marbles and 92 white marbles. From which tray
1843 would you prefer to select a marble in a real situation?
1844 Choice options: the small tray, the large tray
1845
1846 Results:
1847 Wave 1: consistent with Stanovich and West (2008), we find that a significant minority of
1848 participants chose the non-normative large tray ($Prop_{large\ tray} = 0.36$, 95% CI=[0.34,0.38]).
1849 Wave 4: similar results ($Prop_{large\ tray} = 0.38$, 95% CI=[0.36,0.40]).

1850 A3.7 Pricing study (Gui and Toubia 2023)

1851 We replicate the study in Gui and Toubia (2023). See original paper for details. Table A4
1852 lists the set of 40 products in the study. For each product, we vary the price from 0 to
1853 200% of the regular price, in 20% increments. Each respondent answered one purchase
1854 intention question per product, with prices randomly drawn for each product and the order
1855 of products randomized across respondent. The wording of the question was as follows for
1856 product in category at price:
1857 Please consider the following product category: {category}.
1858 Suppose you are in a grocery store, and you see the following product in that category: {product}.
1859 The product is priced at: {price}.
1860 Would you or would you not purchase the product? (yes, I would purchase the product; No, I
1861 would not purchase the product).

Table A4: Categories, products and regular prices (from Gui and Toubia (2023))

Category	Product	Price (\$)
Fruit Juice	Capri Sun Variety Pack with Fruit Punch, Strawberry Kiwi & Pacific Cooler Juice Box Pouches, 30 ct Box, 6 fl oz Pouches	9.43
Fruit Drinks	Kool Aid Jammers Variety Pack with Tropical Punch, Grape & Cherry Kids Drink 0% Juice Box Pouches, 30 Ct Box, 6 fl oz Pouches	7.27
Baby Milk and Milk Flavoring	Horizon Organic Shelf-Stable Whole Milk Boxes, 8 oz., 12 Pack	13.98
Soup	Maruchan Ramen Noodle Chicken Flavor Soup, 3 Oz, 12 Count Shelf Stable Package	9.97
Cat Food - Wet Type	Purina Fancy Feast Chicken Feast Classic Grain Free Wet Cat Food Pate - 3 oz. Can	0.88
Pet Supplies - Dog Food	Purina Dog Chow Complete, Dry Dog Food for Adult Dogs High Protein, Real Chicken, 44 lb Bag	29.17
Snacks - Potato Chips	Lay's Classic Potato Snack Chips, Party Size, 13 oz Bag	5.44
Snacks - Tortilla Chips	Doritos Nacho Cheese Tortilla Snack Chips, Party Size, 14.5 oz Bag	5.94
Cereal - Ready to Eat	Cinnamon Toast Crunch Breakfast Cereal, Crispy Cinnamon Cereal, Family Size, 18.8 oz	4.93
Cookies	Little Debbie Oatmeal Creme Pies, 12 ct, 16.2 oz	2.68
Ground and Whole Bean Coffee	Folgers Classic Roast Ground Coffee, Medium Roast, 40.3-Ounce Canister	13.24
Soft Drinks - Carbonated	Coca-Cola Soda Pop, 12 fl oz, 12 Pack Cans	8.26
Bottled Water	OZARKA Brand 100% Natural Spring Water, 16.9-ounce plastic bottles (Pack of 35)	19.96
Candy - Chocolate	Hershey's Milk Chocolate Candy, Bars 1.55 oz, 6 Count	6.48
Candy - Non-Chocolate	HARIBO Goldbears Original Gummy Bears, 28.8oz Stand Up Bag	6.48
Soft Drinks - Low Calorie	Coca-Cola Zero Sugar Soda Pop, 16.9 fl oz, 6 Pack Cans	5.18
Frozen Italian Entrees	Smart Ones Three Cheese Ziti Marinara Frozen Meal, 9 Oz Box	2.26
Frozen Foods	Great Value All Natural Chicken Wing Sections, 4 lb (Frozen)	12.98
Ice Cream	Haagen Dazs Coffee Ice Cream, Gluten Free, Kosher, 14.0 oz	4.18
Frozen Novelties	Pop-Ice Assorted Fruit Freezer Ice Pops, Gluten-Free Snack, 1.5 oz, 80 Count Fruit Pops	6.17
Lunchmeat - Sliced - Refrigerated	Oscar Mayer Chopped Ham & Water product Deli Lunch Meat, 16 Oz Package	4.33
Frankfurters - Refrigerated	Oscar Mayer Classic Uncured Beef Franks Hot Dogs, 10 ct Pack	3.94
Refrigerated Bacon	Oscar Mayer Fully Cooked Original Bacon, 2.52 oz Box	4.27
Refrigerated Entrees	John Soules Foods Chicken Breast Fajita Strips, Refrigerated, 16oz, 18g Protein per 3oz Serving Size	5.98
Dairy Products	Land O Lakes Salted Stick Butter, 16 oz, 4 Sticks	5.28
Yogurt - Refrigerated	Chobani Non-Fat Greek Yogurt, Vanilla Blended 32 oz, Plastic	5.58
Refrigerated Deli Meats	Goya Cooked Ham 16 oz	29.99
Dairy - Milk - Refrigerated	Great Value Milk Whole Vitamin D Gallon	3.92
Bakery - Fresh Cakes	Little Debbie Zebra Cakes, 13 oz	2.68
Fresh Eggs	Eggland's Best Classic Extra Large White Eggs, 12 count	3.18
Fresh Fruit	Fresh Raspberries, 12 oz Container	4.74
Beer	Stella Artois Lager, 12 Pack, 11.2 fl oz Glass Bottles, 5% ABV, Domestic Beer	15.73
Light Beer (Low Calorie/Alcohol)	Bud Light Beer, 24 Pack, 12 fl oz Aluminum Cans, 4.2% ABV, Domestic Lager	20.98
Detergents - Heavy Duty - Liquid	Purex Liquid Laundry Detergent Plus OXI, Stain Defense Technology, 128 Fluid Ounces, 85 Wash Loads	9.97
Cleaning Supplies	ARM & HAMMER Pure Baking Soda, For Baking, Cleaning & Deodorizing, 1 lb Box	1.54
Toilet Tissue	Angel Soft Toilet Paper, 9 Mega Rolls, Soft and Strong Toilet Tissue	6.68
Paper Towels	Bounty Select-a-Size Paper Towels, 12 Double Rolls, White	22.18
Batteries	Duracell Coppertop AA Battery, Long Lasting Double A Batteries, 16 Pack	15.97
Pain Remedies - Headache	Tylenol Extra Strength Caplets with 500 mg Acetaminophen, 100 Ct	10.97
Cold Remedies - Adult	Equate Value Size Honey Lemon Cough Drops with Menthol, 160 Count	4.68