

Memory as a Belief Technology

Pedro Bordalo (Oxford) Nicola Gennaioli (Bocconi)
Matthew O'Brien (LSE) Andrei Shleifer (Harvard)

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Motivation: Experiences Shape Beliefs

Intuition: Past experiences are a building block for what we think will happen in the future: If the last week was sunny every day, I might expect tomorrow will also be sunny also.

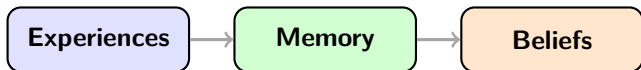


“Experience effects” papers:

- **Malmendier & Nagel (2011):** Individuals who grew up during the Great Depression are more pessimistic about future returns and invest less in stocks for the rest of their lives → formative macroeconomic experiences have large, persistent effects on financial decisions.
- **Malmendier & Shen (2024):** Households that lived through recessions persistently think a future recession is more likely and under consume relative to their fundamentals, even decades after recovery.

Motivation: Experiences Operate *Through* Memory

Intuition: if you cannot remember last week's sunny days, those experiences are unlikely to shape your belief about tomorrow's weather.



Memory determines how well experiences are accessed:

- **Tulving (2002):** Episodic memory underlies “mental time travel.” Amnesiac patients cannot simulate future events or form stable forward-looking beliefs.
- **Klein, Loftus & Kihlstrom (2002):** Patients with amnesia cannot construct or update self-relevant beliefs about the future; their beliefs are highly unstable even across short intervals.

Memory-Based Models of Belief Formation

- Motivates a class of "Memory-Based Models" of Belief Formation (Bordalo et al., 2023; Wachter & Kahana, 2024).

■ Model Setup

- Individuals have a database of experiences $E = \{e_1, e_2, \dots\}$
- Each experience has contextual features $F = \{f_1, \dots, f_k\}$
- Similarity $S(e, H)$ between features governs recall probability $r(e)$

■ Belief Decomposition

$$\underbrace{\widehat{\pi(H)}}_{\text{P(event)}} = \sum_{e \in E} \underbrace{r(e)}_{\text{recall prob}} \cdot \sigma(e) \quad \text{where} \quad r(e) = \frac{S(e, H)}{\sum_{u \in E} S(u, H)}$$

- Probability judgments emerge from *similarity-based retrieval* of past experiences.
- "Irrelevant" information can influence beliefs if it shares contextual features with the event H (Bordalo et al., 2024)

Empirical Tests of Memory-Based Models

- **Bordalo, Gennaioli & Shleifer (2024, Restud)** — “**Imagining the Future**”: People assess novel risks by simulating them from similar past experiences rather than from base rates, causing personal history to systematically bias beliefs.
- **Enke, Schwerter & Zimmermann (2024, JFE)** — “**Associative Memory**”: When people receive new information, they selectively recall past news that shares the same contextual features.
- **Jiang, Liu, Peng & Yan (2025, QJE)** — “**Investor Memory and Biased Beliefs**”: Investors’ recalled (not actual) past experiences drive return expectations — rising markets trigger selectively positive memories, leading to excess optimism.
- **Andre, Pizzinelli, Roth & Wohlfart (2022, Restud)** — “**Subjective Models of the Macroeconomy**”: Irrelevant cues can shift mental models for how macro shocks will propagate.

This Paper

The Gap: Existing work treats memory as a black box. Without measuring individual memory strength, we cannot know *by how much* memory moderates the role of experiences (if at all).

This paper:

- 1 Directly measure memory strength** (Recall Fluency, RF) using validated cognitive tests.
- 2 Test whether RF amplifies experience effects on beliefs:** We show that high-RF individuals have significantly larger "experience effects" compared to low-RF.
- 3 Memory as a stable trait:** like risk aversion or IQ, RF is largely stable within-person over time — a fundamental, persistent driver of belief formation.

Setting: Stock market beliefs (chance stocks \uparrow next year), using each individual's lifetime S&P 500 return history as experiences.

A Model of Belief Formation With Memory Strength

(1) Setup

- **Database** $\mathbf{f} = \{f_e\}_{e \in E}$: lifetime experiences stored as frequency counts over categories $e = (s, v, c)$ — signal, valence, context.
- **Cue** $q_X = (s_X, v, X)$: current signal, mood, and domain. Can think of q_X as the belief question itself.

(2) **Retrieval** — the probability of recalling experience category e given cue q :

$$r_e(q, \mathbf{f}) = \frac{S(e, q) \cdot f_e}{\sum_{e'} S(e', q) \cdot f_{e'}}$$

- *Frequency*: more commonly lived experiences are easier to recall
- *Similarity*: experiences closer in features (s, v, c) to the cue are easier to recall
- *Interference*: Irrelevant experiences crowd out relevant ones

Recall Fluency: From Retrieval to Beliefs

(3) Attention a_i governs similarity gradient: $S(e, q) = S[\alpha_i \cdot d(e, q)]$

- Higher $a_i \Rightarrow$ steeper gradient \Rightarrow cue-congruent experiences upweighted, irrelevant experiences suppressed.

(4) Belief Decomposition:

$$\mathbb{E}(s_X|q, \mathbf{f}) = \sum_{e: \sigma_X(e) \in T_X} \underbrace{r_e(q, \mathbf{f})}_{\text{recall prob}} \cdot \underbrace{\sigma_X(e)}_{\text{simulation}}$$

Linearizing around $a_i = 0$ yields the belief decomposition:

$$\mathbb{E}(s_X|q, \mathbf{f}) \approx \underbrace{\sum_e \frac{f_e}{|\mathbf{f}|} \sigma_X(e)}_{\text{frequency anchor}} + \underbrace{a_i \sum_e \frac{f_e}{|\mathbf{f}|} [d(q) - d(e, q)] \sigma_X(e)}_{\text{RF amplification of similar experiences}}$$

Reduced Form Regression

Empirical Analogue:

$$p_{X,it} \approx \eta_0 + \underbrace{\eta_1 \cdot \bar{s}_{it}}_{\text{frequency anchor}} + \underbrace{\eta_2 \cdot a_i \cdot \bar{s}_{it}}_{\text{RF} \times \text{avg. experience}} - \underbrace{\eta_3 \cdot a_i \cdot \widehat{\text{COV}}_i}_{\text{RF} \times \text{similarity to today}}$$

- \bar{s}_{it} : average past DS experience in domain X — the frequency anchor on beliefs.
- $\eta_2 > 0$: high RF amplifies sensitivity to the average past experience database.
- $\widehat{\text{COV}}_i \equiv \text{cov}[s, (s - s_{it})^2]$: how closely past experiences s resemble today's signal s_{it} .
- $\eta_3 > 0$: high RF further amplifies recall of experiences *similar to today* — the same news hits differently depending on whether it matches your personal history.

Testable Predictions

Prediction 1 — Agnosticity: Low-RF individuals give more *agnostic* answers: they are more likely to report “Don’t Know” or an uncertain 50/50 when asked about future returns.

- High-RF individuals are more able to commit to a directional belief.

Prediction 2 — Database (DS) Amplification:

- $\hat{\eta}_2 > 0$: High-RF individual who has experienced good market returns is more bullish than a low-RF individual with the same experiences.
- $\hat{\eta}_3 < 0$: High-RF individuals disproportionately recall past experiences similar (low $cov[s, (s - s_{it})^2]$) to today’s signal — the same news hits differently depending on whether it rhymes with your personal history.

Prediction 3 — Non-Database (NDS) Experiences: RF also amplifies the role of experiences *outside* the financial database.

- Early-life episodes stored in memory that shape the context for later recall (childhood health, childhood SES)

Health and Retirement Study (HRS): nationally representative panel of Americans aged 50+, surveyed biennially 2002–2020 ($\approx 20,000$ individuals)

Belief variables (outcomes):

- Chance stocks \uparrow [0-100]
- Chance house prices \uparrow [0-100]
- Chance survive 10+ years [0-100]

Experience variables:

- \bar{s}_{it} : mean lifetime S&P 500 return
- $\widehat{\text{cov}}_{it}$: $\text{cov}[s, (s - s_{it})^2]$
- Childhood health index (NDS)
- Childhood SES index (NDS)

Cognition measures:

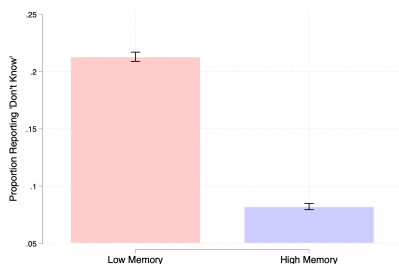
- **RF**: standardized index of immediate recall, delayed recall, verbal fluency (memory strength)
- **IQ**: math index (numeracy, counting backwards, series)

Controls:

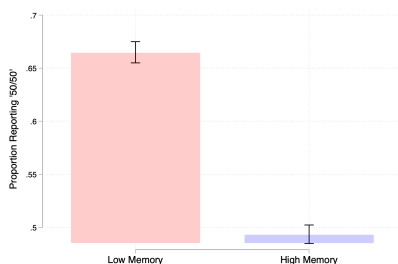
- Age, gender, marital status, retirement, income, education

Prediction 1: Low RF \Rightarrow More Agnostic Answers

Prop. "Don't Know", Stocks



Prop. "50/50 + Unsure", Stocks



- **Takeaway:** Low-RF individuals are significantly more likely to say "Don't Know" (left) and, among those who answer 50/50, more likely to report genuine uncertainty (right).

Prediction 2: RF Amplifies Experience Effects

	<u>Chance Stocks</u> ↑	
	(1)	(2)
Cognition		
RF	.	0.002 [0.005]
DS Experience		
\bar{s}_{it} ($\hat{\eta}_1$)	0.001 [0.002]	0.002 [0.003]
RF Interactions		
RF × \bar{s}_{it} ($\hat{\eta}_2$)	0.006*** [0.000]	0.006*** [0.002]
RF × \widehat{cov}_i ($\hat{\eta}_3$)	-0.004*** [0.000]	-0.004*** [0.000]
N	139345	139345
DV Mean	0.47	0.47

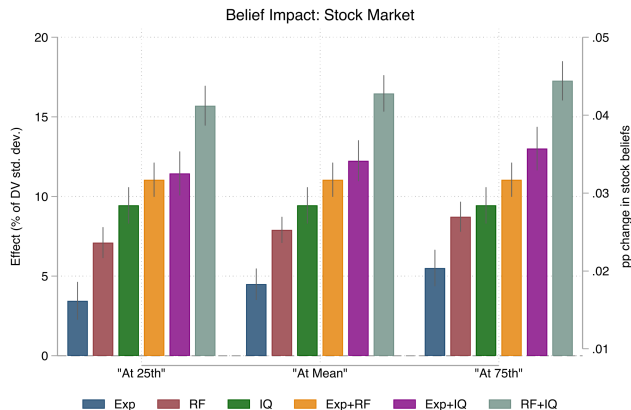
Takeaways:

$\hat{\eta}_1 > 0^{**}$: Lifetime experiences anchor beliefs — more positive past returns \Rightarrow more optimistic.

$\hat{\eta}_2 > 0$: RF amplifies this — high-RF individuals are more sensitive to their personal history.

$\hat{\eta}_3 < 0$: RF further amplifies recall of experiences *similar to today* — same news hits differently depending on personal history.

Prediction 2: The Role of Memory — Graphical



- **Takeaway** RF alone explains a meaningful share of belief volatility. 1 std change in RF worth twice as much as 1 std change in experiences.

Prediction 3: Memory Amplifies Early-Life Experiences

	Chance Stocks \uparrow	
	(1) Child. Health	(2) Child. SES
DS Experience		
\bar{s}_{it} ($\hat{\eta}_1$)	0.001 [0.002]	0.001 [0.002]
RF Interactions		
RF \times \bar{s}_{it} ($\hat{\eta}_2$)	0.006*** [0.000]	0.006*** [0.000]
RF \times \widehat{cov}_i ($\hat{\eta}_3$)	-0.003*** [0.000]	-0.004*** [0.000]
NDS Experience		
NDS	-0.001 [0.001]	-0.003 [0.002]
RF \times NDS	0.005*** [0.001]	0.005*** [0.001]
N	134136	139342
DV Mean	0.47	0.47

Takeaways:

- High-RF individuals carry childhood health and SES into adult beliefs via memory's amplification.
- Happens through "valence-based" channel of the experience $e = (s,v,c)$
- NDS only matter *through* RF — consistent with a memory channel, not omitted variables.
- **Memory makes even normatively irrelevant experiences matter**

Taking Stock

	Prediction	Result
P1	<i>Agnosticity</i> : Low RF \Rightarrow more agnostic answers (DK or uncertain 50/50)	✓ Low-RF more likely to report “Don’t Know” or uncertain 50/50
P2	<i>DS amplification</i> : RF amplifies lifetime market experience effects	✓ $\hat{\eta}_2 > 0$: RF $\times \bar{s}_{it}$ significant; $\hat{\eta}_2 < 0$: RF $\times cov[s, (s - s_{it})^2]$
P3	<i>NDS amplification</i> : RF amplifies early-life experience effects	✓ RF \times Child Health and RF \times Child SES impact beliefs <i>through memory</i>

All three predictions of the memory-based model are supported in the data.

Memory actively mediates the relationship between experience and belief — and explains more belief variance than lifetime experiences alone.

Conclusion

- **Memory as a belief technology:** we propose and test a memory-based model of belief formation; the data are strongly consistent across three distinct predictions.
- **Memory is a stable individual trait [not discussed today]** — like risk aversion or IQ — that varies substantially across people, creating a persistent wedge between the experiences people have and the beliefs they form.
- **Memory amplifies experience effects** — for both lifetime market returns (database) and early-life conditions (non-database). Individuals with identical experience histories form different beliefs based on cognitive endowment.
- **Memory explains more belief variance** than lifetime experiences alone (in the stock market), underscoring its first-order importance for understanding cross-sectional heterogeneity in beliefs and financial decisions.

Robustness: Housing Market and Health Beliefs

	<u>Chance House ↑</u>		<u>Chance Live 10+</u>	
	(1)	(2)	(3)	(4)
<i>Cognition</i>				
RF	.	-0.013 [0.010]	.	0.006** [0.003]
IQ	0.016*** [0.003]	0.017*** [0.003]	-0.018*** [0.002]	-0.019*** [0.002]
<i>DS Experience</i>				
\bar{s}_{it} ($\hat{\eta}_1$)	0.038*** [0.003]	0.029*** [0.008]	0.001*** [0.000]	0.001*** [0.000]
<i>RF Interactions</i>				
RF \times \bar{s}_{it} ($\hat{\eta}_2$)	0.006*** [0.001]	0.011*** [0.004]	0.000*** [0.000]	0.000*** [0.000]
RF \times $\widehat{\text{cov}}_i$ ($\hat{\eta}_3$)	-0.035*** [0.001]	-0.035*** [0.001]	-0.000*** [0.000]	-0.000*** [0.000]
N	49972	49972	129518	129518
Adj. R ²	0.053	0.053	0.102	0.102
DV Mean	0.53	0.53	0.47	0.47

Notes: Same specification as Slide 10. Cols (1)–(2): housing market beliefs; cols (3)–(4): health beliefs (chance live 10+ years). Memory amplification of experience effects is robust across all three belief domains.