

Habits as Adaptations: An Experimental Study

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jointly with

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ES European Winter Meeting, Naples
December 4, 2018

Introduction

How do people respond to changing incentives?

Puzzling behavior: choice inertia

- ▶ inertia in consumption
- ▶ brand loyalty
- ▶ status quo bias

Habits in macroeconomics

Assuming *preferences for habits*

$$u(c_t - c^{t-1})$$

Justification: Better fit to data (e.g. Constantinides 1990; Fuhrer, 2000)

Problems:

1. No microfoundations
evidence for inertia \neq evidence for preferences for habits
2. Modeling choice of c^{t-1} not obvious
 - ▶ aggregate past consumption, past individual consumption, specific categories of goods (Schmitt-Grohé and Uribe, 2007)
3. No comparative statics predictions

This paper: Testing microfoundations of choice inertia

Microfoundations of choice inertia

Habits in psychology:

- ▶ “Automated responses triggered by *cues* to alleviate *cognition costs*” (e.g. Lally et al., 2010)
- ▶ Cue = element from history which (empirically) correlates with optimal current choice

Research questions

1. Do habits arise to save on cognition/information costs?
2. How are cues selected?
 - ▶ Mechanically?
 - ▶ In a predicted way (optimally)?

⇒ a model of costly information acquisition (Steiner, Steward, and Matějka, 2017)

Preview

One binary perception task in each of two periods

Time separable utility

Treatments:

1. Underlying stochastic process and stakes
⇒ impacts **whether** habit arises and its **strength**
2. Information provided to subjects
⇒ impacts **cue selection**

Summary:

Habits and cue selection as second-best adaptations

Literature

Theory

Cue-theory based on psychology: Laibson (2001)

Neuroscience: Camerer, Landry and Webb (2018)

Rational inattention: Steiner, Steward, Matějka (2017)

Experiment

Khaw and Zorilla (2018)

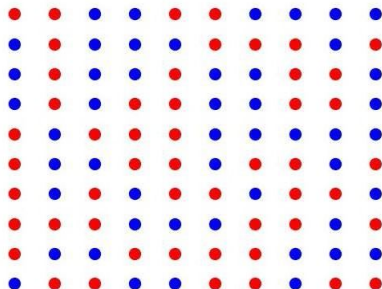
Outline

Experimental design

Hypotheses

Results

Caplin and Dean (2014)



- ▶ 100 red and blue dots
- ▶ Two states: 49 red dots vs. 51 red dots
- ▶ Task: determine the predominant color
 - ▶ cognitive cost \Rightarrow cost-benefit analysis

Our paper: Two periods

Timing

Two-period task

1. State at $t = 1$ drawn from uniform prior

Timing

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2. Subjects make choice at $t = 1$

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4. State at $t = 2$ drawn (known correlation)

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2. Subjects make choice at $t = 1$
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5. Subjects make choice at $t = 2$

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6. Correct answers for both periods revealed

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One treatment = 12 iterations of this two-period task

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Treatments vary alongside

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1. Information feedback: point 3.

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2. Serial correlation (correlated/independent): point 4.

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Treatments vary alongside

1. Information feedback: point 3.
2. Serial correlation (correlated/independent): point 4.
3. Stakes and degree of correlation

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Definitions

Definition: Habit

DM forms a **habit** if payoff-irrelevant elements of history predict continuation behavior.

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DM forms a **habit** if

$$\Pr[a_2 = 1 | a_1, \theta_1, \theta_2] \neq \Pr[a_2 = 1 | \theta_2].$$

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$$\Pr[a_2 = 1 | a_1, \theta_1, \theta_2] \neq \Pr[a_2 = 1 | \theta_2].$$

Which of $\{a_1, \theta_1\}$?

Definitions

Definition: Cues

$z \in \{a_1, \theta_1\}$ is the **cue** for the habit if

1. the habit is solely triggered by z and not by both past variables, and
2. z and a_2 are positively 'correlated'.

Definitions

Definition: Cues

$z \in \{a_1, \theta_1\}$ is the **cue** for the habit if

1. $\Pr[a_2 = 1 | \theta_2, z, w] = \Pr[a_2 = 1 | \theta_2, z]$ and
2. $\Pr[a_2 = 1 | \theta_2, z = 1] > \Pr[a_2 = 1 | \theta_2, z = 0]$

where w is the complementary variable from $\{\theta_1, a_1\}$.

Hypotheses

Based on a theoretical model of rational inattention

- ▶ model of costly information acquisition

Specific case of Steiner, Steward, and Matějka (2017)

Augmented for definition of habit, cue selection, and habit strength

Predictions captured in two propositions

Hypotheses

Weak treatments (W)

<i>high stake ($s = \\$10$)</i>	<i>no feedback (N)</i>	<i>feedback (F)</i>
<i>independent (I)</i>	no habit	no habit
<i>correlation (C)</i> <i>low ($\gamma = 0.75$)</i>	weak habit cue a_1	weak habit cue θ_1

Strong treatments (S)

<i>low stake ($s = \\$7$)</i>	<i>no feedback (N)</i>	<i>feedback (F)</i>
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Correlation \Rightarrow habits as a way to save on costly effort

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Which cue? \Rightarrow depends on its information content

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How strongly? \Rightarrow depends on cost and probability of possible mistakes

Outline

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Results

Data: Overview

University of California, Santa Barbara

4 sessions, 76 subjects

- ▶ 2 sessions 'Weak' treatments, 2 sessions 'Strong' treatments

96 decision problems per subject

492 observations per period

\$10 show-up fee, \$10 vs. \$7 incentive

Data: basic statistics

Treatment \ Frequency	$a_1 = \theta_1$	$a_2 = \theta_2$	$a_2 = a_1$	$\theta_2 = a_1$
<i>INW</i>	0.84	0.86	0.51	0.50
<i>IFW</i>	0.85	0.85	0.60	0.61
<i>CNW</i>	0.87	0.86	0.78	0.74
<i>CFW</i>	0.89	0.90	0.78	0.77
<i>INS</i>	0.87	0.85	0.51	0.50
<i>IFS</i>	0.82	0.82	0.53	0.55
<i>CNS</i>	0.84	0.85	0.91	0.84
<i>CFS</i>	0.86	0.87	0.75	0.73

Independent (I) vs. Correlated (C)

Feedback (F) vs. No Feedback (N)

Weak (W) vs. Strong (S) habit

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- ▶ heterogeneous on the individual level

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- ▶ aggregate accuracy high and homogeneous across treatments and periods
- ▶ heterogeneous on the individual level
- ▶ hard to extract much more information just by looking at the table

Logit regressions: separately for each treatment

LHS: $a_{2,i}^n$

RHS: $const.$, $a_{1,i}^n$, θ_1^n , θ_2^n , $session$, $score_i^n$, $score_i^n \theta_2^n$

a_t action at $t = 1, 2$

θ_t state at $t = 1, 2$

$score$ (adjusted) total number of correct answers

$session$ a session dummy

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Interested in how θ_1 and a_1 predict a_2 .

Results: habit formation and cue selection

	IFW	INW	CFW	CNW
a_1	-.021 (.036)	.034 (.041)	.017 (.032)	.191*** (.051)
θ_1	.071 (.043)	-.026 (.049)	.258*** (.058)	.002 (.036)
θ_2	.681*** (.032)	.692*** (.054)	.611*** (.046)	.629*** (.067)
	IFS	INS	CFS	CNS
a_1	-.031 (.037)	.037 (.045)	-.033 (.204)	.511*** (.110)
θ_1	.009 (.040)	-.034 (.044)	.498*** (.098)	
θ_2	.632*** (.045)	.700*** .036	.425*** (.121)	.367*** .098

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3. Correlated states \Rightarrow habits

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Results: habit strength

Pooling data across weak and strong parameters for correlated treatments

- ▶ dummy variable $\delta \in \{0, 1\}$

Augmented regression: adding δ and its interaction terms

Estimation of $\Delta(\text{cue}) :=$ marginal effect of cue when $\delta = 1$ (S) minus marginal effect of cue when $\delta = 0$ (W)

Findings

1. No feedback treatments: $\Delta(a_1) = 0.314$ with p-value 0.009
2. Feedback treatments: $\Delta(\theta_1) = 0.234$ with p-value 0.06

Conclusion

Laboratory experiment to test habit formation

Findings consistent with the second-best behavior

Complementary question: Internalizing continuation value of information?

- ▶ Myopia?