# Does Saving Cause Borrowing? Implications for the Co-Holding Puzzle 

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## Summary of paper

- Question: Why do people co-hold high-cost credit card debt and lowinterest liquid savings?
- Empirical setting: a controlled experiment of SMS saving nudges by bank
- Theoretical derivations for the effect of such saving nudges on spending, savings and debt under (a) liquidity premium vs (b) mental accounting models.
- Causal forest approach to estimate individual-level treatment effect
- Main findings:
- Substantial heterogeneity across individuals
- Among those who respond to the saving nudges, spending $\downarrow$, savings $\uparrow$ while credit card debt stay unchanged $\rightarrow$ supports the mental accounting predictions


## Outline of discussion

- Fantastic paper that I enjoy reading:
- Clear theoretical predictions
- Excellent empirical execution
- Novel insights for classic question from methodological innovations
- Comments and ideas
- Using the vast heterogeneity to understand different mechanisms
- Comparing and interpreting different nudges
- Implications for designing and analyzing behavioral interventions


## Co-holding

- Simultaneously holding high-interest revolving debt and low-yielding liquid assets
- Three features of co-holding across contexts and definitions
- Prevalent
- Costly
- Persistent
- A puzzle? A mistake?


## A classic question: Why do people co-hold?

- (Rational) inattention
- Strategic option ahead of bankruptcy
- Insurance against risk that credit limit is reduced
- Emergency savings
- Self-control
- Mental accounting
- Payment preferences


## Theoretical predictions for spending, savings, and debt

Model and effect of a saving nudge Spending Savings Debt

Liquidity premium model

| Patience $\uparrow$ | $\downarrow$ | $\sim$ | $\downarrow$ |
| :---: | :---: | :---: | :---: |
| Liquidity need $\uparrow$ | $\sim$ | $\uparrow$ | $\uparrow$ |
| Mental accounting model |  |  |  |
| Patience $\uparrow$ | $\downarrow$ | $\uparrow$ | $\sim$ |

## Overall effects support mental accounting predictions

|  | Ln Spending +1 | Ln Checking <br> Account <br> Balance +1 | Ln Credit Card Interest +1 During Treat. | Paid Interest During Treat. $\{0,1\}$ | Ln Ending Statement Balance <br> - Payments After Treat. +1 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: All Individuals |  |  |  |  |  |
| Any Treatment | $\begin{aligned} & -0.009^{*} \\ & (0.005) \end{aligned}$ | $\begin{aligned} & 0.006^{*} \\ & (0.003) \end{aligned}$ |  |  |  |
| Observations | 3,054,503 | 3,054,503 |  |  |  |
| Mean of Dep.Var. in Control Group | 16,732.41 | 17,393.63 |  |  |  |
| Panel B: Individuals with a Credit Card |  |  |  |  |  |
| Any Treatment | $\begin{gathered} -0.021^{* * *} \\ (0.006) \end{gathered}$ | $\begin{aligned} & 0.012^{* *} \\ & (0.006) \end{aligned}$ | $\begin{aligned} & -0.004 \\ & (0.004) \end{aligned}$ | $\begin{gathered} -0.001 \\ (0.004) \end{gathered}$ | $\begin{aligned} & -0.003 \\ & (0.005) \end{aligned}$ |
| Observations | 362,223 | 362,223 | 362,223 | 362,223 | 362,223 |
| Mean of Dep.Var. in Control Group | 29,960.75 | 34,586.21 | 213.84 | 0.41 | 4,981.45 |
| Panel C: Individuals with a Credit Card Who Paid Interest at Baseline |  |  |  |  |  |
| Any Treatment | $\begin{gathered} -0.019^{* *} \\ (0.007) \end{gathered}$ | $\begin{aligned} & 0.017^{* *} \\ & (0.007) \end{aligned}$ | $\begin{gathered} -0.004 \\ (0.005) \end{gathered}$ | $\begin{gathered} -0.001 \\ (0.005) \end{gathered}$ | $\begin{gathered} 0.002 \\ (0.005) \end{gathered}$ |
| Observations | 152,016 | 152,016 | 152,016 | 152,016 | 152,016 |
| Mean of Dep.Var. in Control Group | 31,818.77 | 31,940.83 | 479.14 | 0.81 | 10,219.67 |

## Substantial heterogeneity across individuals



## Roughly $1 / 4$ respond in spending, $1 / 4$ respond in savings



## Joint distribution of the two dimensions is revealing

(c) Individuals with a Credit Card Who Paid Interest at Baseline

|  | 1 | 2 | 3 | 4 |
| :--- | :---: | :---: | :---: | :---: |
| 1 | 0.1017 | 0.0576 | 0.0571 | 0.0332 |
| 2 | 0.0592 | 0.0901 | 0.0526 | 0.0478 |
| 3 | 0.0571 | 0.0669 | 0.1070 | 0.0195 |
| 4 | 0.0319 | 0.0355 | 0.0339 | 0.1492 |

Spending $\downarrow$
Savings $\uparrow$

## Joint distribution of the two dimensions is revealing

## (c) Individuals with a Credit Card Who Paid Interest at Baseline



If debt $\downarrow$, these $10 \%$ of people look like agents with patience $\uparrow$ under liquidity premium models.

## Comparing different nudges

| Message | Type? | Effect for <br> spending | Effect for <br> savings |
| :--- | :--- | :--- | :--- | :--- |
| 1: "Congratulations. Your average balance over the last 12 months has <br> been great! Continue to increase your balance and strengthen your <br> savings." | Savings in <br> general | -0.0351 | 0.0209 |
| 2: "Increase the balance in your Banorte Account and get ready today <br> for year-end expenses!" | Short-term <br> goals | $-0.0874^{* * *}$ | $0.0516^{* *}$ |
| 3: "Join customers your age who already save 10\% or more of their <br> income. Commit and increase the balance in your Banorte Account by <br> \$XXX this month." | Savings in <br> general | $-0.1216^{* * *}$ | $0.0779^{* * *}$ |
| 4: "In Banorte, you have the safest money box! Increase your account <br> balance by \$XXX this payday and reach your goals." | Mental <br> accounting | $-0.1239^{* * * *}$ | $0.0811^{* * *}$ |
| 5: "Increase your balance this month by \$XXX and reach your dreams. <br> Commit to it. You can do it by saving only 10\% of your income." | Savings in <br> general | $-0.0685^{* * * *}$ | 0.0371 |
| 6: "The holidays are coming. Commit to saving \$XXX in your Banorte <br> Account and avoid money shortfalls at year-end!" | Short-term <br> goals | $-0.0413^{*}$ | 0.0219 |
| 7: "Be prepared for an emergency! Commit to leaving 10\% more in <br> your account. Don't withdraw all your money on payday." | Short-term <br> goals | $-0.0918^{* * *}$ | $0.0546^{* *}$ |

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| 3: "Join customers your age who already save 10\% or more of their <br> income. Commit and increase the balance in your Banorte Account by <br> \$XXX this month." | Peer <br> effects? | $-0.1216^{* * *}$ | $0.0779^{* * *}$ |
| 4: "In Banorte, you have the safest money box! Increase your account <br> balance by \$XXX this payday and reach your goals." | Mental <br> accounting | $-0.1239^{* * *}$ | $0.0811^{* * *}$ |
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## Big picture: designing and analyzing interventions

- The causal forest approach enables an estimate of treatment effect for each individual without suffering from over-fitting, curse of dimensionality, spurious correlation, or ad-hoc parametric choice for treatment effect heterogeneity
- It reveals that close to $65 \%$ of the treated individuals do not change spending or savings in response to the saving nudges.
- Equally important to understand for whom an intervention works vs does not work.
- Will be interesting to analyze and interpret the characteristics of nonresponders.

