# STICKING TO YOUR PLAN: HYPERBOLIC DISCOUNTING AND CREDIT CARD DEBT PAYDOWN

Theresa Kuchler

April 2014

## GOAL AND DATA

**Question**: Why do people hold large credit card balances at high cost over substantial time horizons?

1

#### Goal and Data

**Question**: Why do people hold large credit card balances at high cost over substantial time horizons?

Goal: Estimate the role of present bias in debt paydown

1

#### GOAL AND DATA

**Question**: Why do people hold large credit card balances at high cost over substantial time horizons?

Goal: Estimate the role of present bias in debt paydown

Data: Sample of users of financial management website

- Account balances
- Spending on credit and debit card
- Income
- Planned paydown

## PRESENT BIAS - DEFINITION

#### **Present Bias**

• More impatient in short-run  $(\beta \delta)$  than long-run  $(\delta)$ 

$$U_t = u(c_t) + \beta \sum_{\tau=t+1}^{\infty} \delta^{\tau} u(c_{\tau})$$

→ Time inconsistent

## Present Bias - Definition

#### **Present Bias**

• More impatient in short-run  $(\beta \delta)$  than long-run  $(\delta)$ 

$$U_t = u(c_t) + \beta \sum_{\tau=t+1}^{\infty} \delta^{\tau} u(c_{\tau})$$

- → Time **in**consistent
- Two present-biased types
  - Fully Aware → Sophisticated
  - Unaware of changing preferences → Naive

#### Present Bias - Definition

#### **Present Bias**

• More impatient in short-run  $(\beta\delta)$  than long-run  $(\delta)$ 

$$U_t = u(c_t) + \beta \sum_{\tau=t+1}^{\infty} \delta^{\tau} u(c_{\tau})$$

- → Time **in**consistent
- Two present-biased types
  - Fully Aware → Sophisticated
  - Unaware of changing preferences → Naive
  - $\rightarrow$  Two features of present bias:
    - Extent of short-run impatience  $(\beta)$
    - Sophistication

1. Infer features of present bias from spending patterns

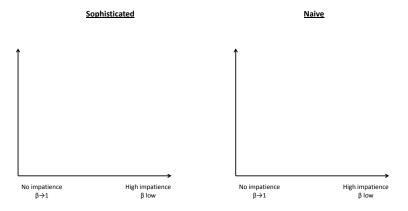
- 1. Infer features of present bias from spending patterns
  - Level of short-run impatience
    - ightarrow Sensitivity of consumption spending to paycheck receipt

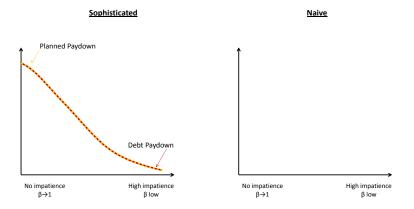
- 1. Infer features of present bias from spending patterns
  - Level of short-run impatience
    - → Sensitivity of consumption spending to paycheck receipt
  - Awareness of future short-run impatience/sophistication
    - → Effect of varying resources on spending patterns

- 1. Infer features of present bias from spending patterns
  - Level of short-run impatience
    - ightarrow Sensitivity of consumption spending to paycheck receipt
  - Awareness of future short-run impatience/sophistication
    - → Effect of varying resources on spending patterns
- 2. Assess who sticks to plan of debt reduction

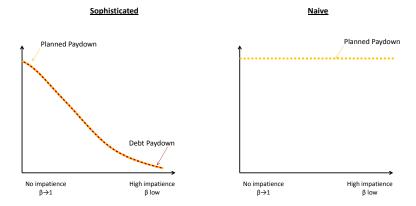
- 1. Infer features of present bias from spending patterns
  - Level of short-run impatience
    - → Sensitivity of consumption spending to paycheck receipt
  - Awareness of future short-run impatience/sophistication
    - → Effect of varying resources on spending patterns
- 2. Assess who sticks to plan of debt reduction
  - Naive/Unaware of future present bias:
    - Repeatedly fail to pay down debt as planned

- 1. Infer features of present bias from spending patterns
  - Level of short-run impatience
    - → Sensitivity of consumption spending to paycheck receipt
  - Awareness of future short-run impatience/sophistication
    - → Effect of varying resources on spending patterns
- 2. Assess who sticks to plan of debt reduction
  - Naive/Unaware of future present bias:
     Repeatedly fail to pay down debt as planned
  - Sophisticated:
    - Follow plan, but pay down less the more impatient

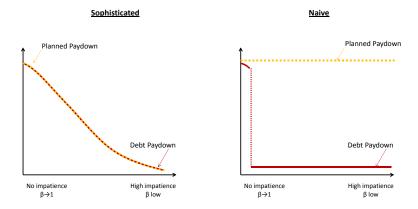




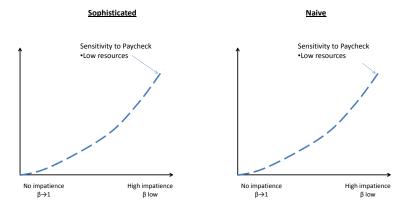
Sophisticated:
 Follow plan to pay down
 Pay down less the more impatient

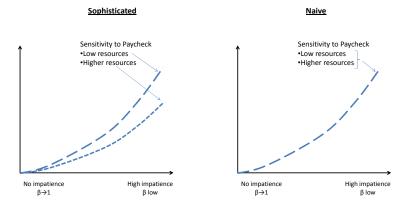


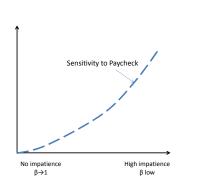
• Naive:



 Naive: Repeatedly plans to pay off next period

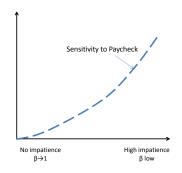


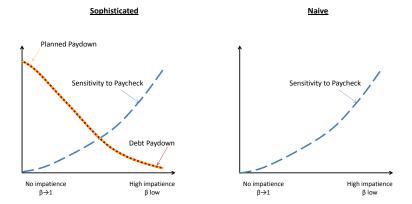




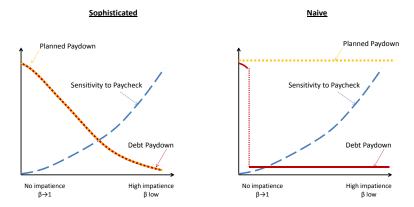
Sophisticated

#### **Naive**





• *Sophisticated*:
Pay down less the more impatient



- Sophisticated:
   Pay down less the more impatient
- Naive:
   Delays irrespective of plans and impatience level

#### REGRESSION RESULTS - DEBT PAYDOWN

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i \mu_{1n} + PlannedPaydown_i \mu_{2n} \\ &+ Sensitivity_i * Sophist_i \mu_{1s} \\ &+ PlannedPaydown_i * Sophist_i \mu_{2s} \\ &+ X_i' \lambda + \nu_i \end{split}$$

#### REGRESSION RESULTS - DEBT PAYDOWN

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i\mu_{1n} + PlannedPaydown_i\mu_{2n} \\ &+ Sensitivity_i*Sophist_i\mu_{1s} + PlannedPaydown_i*Sophist_i\mu_{2s} + X_i'\lambda + \nu_i \end{split}$$

	Paydown 90 Days		Paydown 180 Days	
	Short-run	Restaurant	Short-run	Restaurant
	Consumables	&Entertainment	Consumables	&Entertainment
Sensitivity	7.35	-5.58	5.10	3.11
	(0.12)	(0.28)	(0.13)	(0.42)
Planned Paydown	0.14**	0.14**	0.07	0.09
	(0.02)	(0.04)	(0.19)	(0.10)
Sensitivity ×	-20.41***	-7.32	-12.29***	-11.51**
Sophisticated	(0.00)	(0.38)	(0.01)	(0.02)
$\begin{array}{l} {\sf Planned\ Paydown}\times\\ {\sf Sophisticated} \end{array}$	0.21*	0.17	0.23**	0.27**
	(0.09)	(0.22)	(0.04)	(0.02)
Median Paycheck	2.47***	2.56***	0.92***	0.70***
	(0.00)	(0.00)	(0.00)	(0.01)
Original Debt	0.09*** (0.00)	0.10*** (0.00)	0.09*** (0.00)	0.04 (0.17)
Sophisticated	-5.85	-4.78	-4.72	-5.81*
	(0.12)	(0.27)	(0.12)	(0.06)
Constant	-3.06	-3.70*	-0.35	0.11
	(0.10)	(0.10)	(0.81)	(0.94)
Nr of Individuals	556	510	556	510

P-values of bootstrapped standard errors in parentheses. Significance: \*(p<0.10), \*\*(p<0.05), \*\*\*(p<0.01).

#### REGRESSION RESULTS - DEBT PAYDOWN

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i\mu_{1n} + PlannedPaydown_i\mu_{2n} \\ &+ Sensitivity_i*Sophist_i\mu_{1s} + PlannedPaydown_i*Sophist_i\mu_{2s} + X_i'\lambda + \nu_i \end{split}$$

	Paydown 90 Days		Paydown 180 Days	
	Short-run	Restaurant	Short-run	Restaurant
	Consumables	&Entertainment	Consumables	&Entertainment
Sensitivity	7.35	-5.58	5.10	3.11
	(0.12)	(0.28)	(0.13)	(0.42)
Planned Paydown	0.14**	0.14**	0.07	0.09
	(0.02)	(0.04)	(0.19)	(0.10)
Sensitivity ×	-20.41***	-7.32	-12.29***	-11.51**
Sophisticated	(0.00)	(0.38)	(0.01)	(0.02)
Planned Paydown $\times$	0.21*	0.17	0.23**	0.27**
Sophisticated	(0.09)	(0.22)	(0.04)	(0.02)
Median Paycheck	2.47***	2.56***	0.92***	0.70***
	(0.00)	(0.00)	(0.00)	(0.01)
Original Debt	0.09*** (0.00)	0.10*** (0.00)	0.09*** (0.00)	0.04 (0.17)
Sophisticated	-5.85	-4.78	-4.72	-5.81*
	(0.12)	(0.27)	(0.12)	(0.06)
Constant	-3.06	-3.70*	-0.35	0.11
	(0.10)	(0.10)	(0.81)	(0.94)
Nr of Individuals	556	510	556	510

P-values of bootstrapped standard errors in parentheses. Significance: \*(p<0.10), \*\*(p<0.05), \*\*\*(p<0.01).

#### Patterns of debt paydown are

• Consistent with present bias/hyperbolic discounting

#### Patterns of debt paydown are

- Consistent with present bias/hyperbolic discounting
- *Inconsistent* with alternative explanations:
- Direct relation between reduction in sensitivity and paydown



#### Patterns of debt paydown are

- Consistent with present bias/hyperbolic discounting
- *Inconsistent* with alternative explanations:
- Direct relation between reduction in sensitivity and paydown
- Credit constraints: exclude when likely to be constrained



#### Patterns of debt paydown are

- Consistent with present bias/hyperbolic discounting
- *Inconsistent* with alternative explanations:
- Direct relation between reduction in sensitivity and paydown
- Credit constraints: exclude when likely to be constrained

Habits, non-separabilities

#### Results

#### Patterns of debt paydown are

- Consistent with present bias/hyperbolic discounting
- *Inconsistent* with alternative explanations:
- Direct relation between reduction in sensitivity and paydown
- Credit constraints: exclude when likely to be constrained

- Habits, non-separabilities
- Unrealistic expectations, overoptimism

#### Results

#### Patterns of debt paydown are

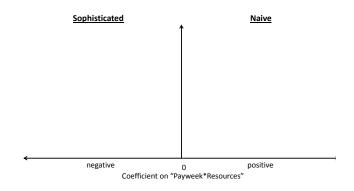
- Consistent with present bias/hyperbolic discounting
- *Inconsistent* with alternative explanations:
- Direct relation between reduction in sensitivity and paydown
- Credit constraints: exclude when likely to be constrained
- Habits, non-separabilities
- Unrealistic expectations, overoptimism
- Impatient users drive difference btw sophisticated and naive



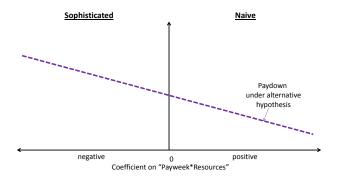
## **Backup Slides**

- Classify based on effect of resources on sensitivity (coefficient on payweek \* resources)
- Higher reduction in sensitivity leads to lower paydown?

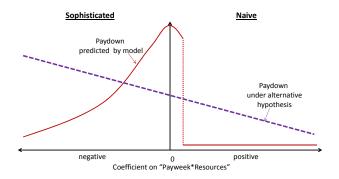
- Classify based on effect of resources on sensitivity (coefficient on payweek \* resources)
- Higher reduction in sensitivity leads to lower paydown?



- Classify based on effect of resources on sensitivity (coefficient on payweek \* resources)
- Higher reduction in sensitivity leads to lower paydown?



- Classify based on effect of resources on sensitivity (coefficient on payweek \* resources)
- Higher reduction in sensitivity leads to lower paydown?



# CLASSIFICATION INTO SOPHISTICATED / NAIVE

 $Paydown_i = \mu_0 + (coefficient\_on\_payweek * resources)\mu_1 + X'_i\lambda + \nu_i$ 

	Paydown 90 Days				Paydown 180 Days				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Short-run Consumable	es								
Coefficient on	0.13	2.32	-0.03	2.03	-0.20	0.88	-0.31	0.66	
Resources*Payweek	(0.83)	(80.0)	(0.96)	(0.13)	(0.67)	(0.25)	(0.52)	(0.39)	
winsorized	1%	5%	1%	5%	1%	5%	1%	5%	
Controls	•		$\checkmark$	$\checkmark$			$\checkmark$	✓	
N	556								
Mean of Regressor	-0.032								
75 <sup>th</sup> - 25 <sup>th</sup> pctile	0.43								

P-values in parentheses. Significance: \* (p<0.10), \*\* (p<0.05), \*\*\* (p<0.01).



## CREDIT CONSTRAINTS

Sensitivity caused by credit constraints?

• Estimation restricted to when spending of payweek affordable in non-payweek

#### CREDIT CONSTRAINTS

### Sensitivity caused by credit constraints?

- Estimation restricted to when spending of payweek affordable in non-payweek
  - → Baseline: *Category spending* possible
  - → Now: 1. *Total discretionary spending* possible
    - Category spending possible with buffer stock (10<sup>th</sup> percentile of resources)

## Total Discretionary Spending Affordable

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i \mu_{1n} + PlannedPaydown_i \mu_{2n} \\ &+ Sensitivity_i * Sophist_i \mu_{1s} + PlannedPaydown_i * Sophist_i \mu_{2s} + X_i' \lambda + \nu_i \end{split}$$

	Paydov	vn 90 Days	Paydown 180 Days			
	Short-run Consumables	Restaurant &Entertainment	Short-run Consumables	Restaurant &Entertainment		
Sensitivity	7.053	-4.861	3.443	3.635		
	(0.179)	(0.388)	(0.324)	(0.288)		
Planned Paydown	0.137*	0.138**	0.068	0.089		
	(0.055)	(0.047)	(0.328)	(0.114)		
${\sf Sensitivity} \times$	-21.645***	-7.184	-10.373**	-12.067**		
Sophisticated	(0.008)	(0.416)	(0.047)	(0.026)		
Planned Paydown $\times$	0.201	0.176	0.216*	0.271**		
Sophisticated	(0.102)	(0.239)	(0.071)	(0.032)		
Sophisticated	-5.715	-5.225	-5.428	-6.283*		
	(0.229)	(0.230)	(0.132)	(0.063)		
Median Paycheck Original Debt Constant	√ √ √	<b>*</b>	<b>√</b> <b>√</b>	<b>*</b>		
Nr of Individuals	551	501	551	501		

P-values of bootstrapped standard errors in parentheses. Significance: \* (p<0.10), \*\* (p<0.05), \*\*\* (p<0.01).

## SPENDING AFFORDABLE WITH BUFFER STOCK

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i \mu_{1n} + PlannedPaydown_i \mu_{2n} \\ &+ Sensitivity_i * Sophist_i \mu_{1s} + PlannedPaydown_i * Sophist_i \mu_{2s} + X_i' \lambda + \nu_i \end{split}$$

	Paydov	vn 90 Days	Paydow	n 180 Days	
	Short-run Consumables	Restaurant &Entertainment	Short-run Consumables	Restaurant &Entertainment	
Sensitivity	-0.157	-13.095***	4.606	2.289	
	(0.974)	(0.009)	(0.156)	(0.527)	
Planned Paydown	0.144**	0.145**	0.060	0.100	
	(0.031)	(0.047)	(0.309)	(0.127)	
${\sf Sensitivity} \times$	-3.684	11.056	-9.248*	-8.057	
Sophisticated	(0.629)	(0.151)	(0.071)	(0.131)	
Planned Paydown $\times$	0.182	0.300**	0.258**	0.299**	
Sophisticated	(0.173)	(0.049)	(0.027)	(0.028)	
Sophisticated	-3.972	-6.586	-6.841**	-7.116*	
	(0.362)	(0.150)	(0.037)	(0.054)	
Median Paycheck Original Debt Constant Nr of Individuals	√ √ √ 542	√ √ √ 490	√ √ √ 542	√ √ √ 490	

P-values of bootstrapped standard errors in parentheses. Significance: \* (p<0.10), \*\* (p<0.05), \*\*\* (p<0.01).

## EXCLUDING LOW SENSITIVITY

Classification into sophisticated/naive of low sensitivity users

- Not meaningful theoretically
- More likely to be potentially misclassified

## EXCLUDING LOW SENSITIVITY

Classification into sophisticated/naive of low sensitivity users

- Not meaningful theoretically
- More likely to be potentially misclassified
- ightarrow Exclude users with low estimated sensitivity/short-run impatience

## Robustness: Impatient users drive results

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i \mu_{1n} + PlannedPaydown_i \mu_{2n} \\ &+ Sensitivity_i * Sophist_i \mu_{1s} + PlannedPaydown_i * Sophist_i \mu_{2s} + X_i' \lambda + \nu_i \end{split}$$

_	Estimates for Short-run Consumables							
		Exclude	e lowest	Exclude lowest				
	Baseline	10%	20%	Baseline	10%	20%		
	90 Days			180 Days				
Sensitivity	7.35 (0.12)	7.18 (0.15)	9.93** (0.04)	5.10 (0.13)	5.46 (0.11)	6.52** (0.05)		
Planned Paydown	0.14** (0.02)	0.14** (0.03)	0.19*** (0.01)	0.07 (0.19)	0.06 (0.26)	0.09 (0.19)		
Sensitivity × Sophisticated	-20.41*** (0.00)	-20.62*** (0.01)	-24.01*** (0.00)	-12.29*** (0.01)	-12.38** -(0.01)	-13.73*** -(0.01)		
Planned Paydown × Sophisticated	0.21* (0.09)	0.21* (0.09)	0.23 (0.14)	0.23** (0.04)	0.23** (0.04)	0.19 (0.17)		
Sophisticated	-5.85 (0.12)	-5.80 (0.15)	-3.97 (0.35)	-4.72 (0.12)	-4.48 (0.14)	-3.03 (0.38)		
Controls								
Median Paycheck Original Debt Constant	<b>√</b> <b>√</b>	✓ ✓	✓ ✓	√ √ √	<b>√</b> <b>√</b>	<b>√ √</b>		
Nr of Individuals	556	500	444	556	500	444		

P-values of bootstrapped standard errors in parentheses. Significance: (p<0.10), \*\* (p<0.05), \*\*\* (p<0.01).

## Robustness: Impatient users drive results

$$\begin{split} Paydown_i &= \mu_0 + Sensitivity_i\mu_{1n} + PlannedPaydown_i\mu_{2n} \\ &+ Sensitivity_i*Sophist_i\mu_{1s} + PlannedPaydown_i*Sophist_i\mu_{2s} + X_i'\lambda + \nu_i \end{split}$$

	Estimates for Restaurant&Entertainment							
	Exclude lowest			Exclude lowest				
	Baseline	10%	20%	Baseline	10%	20%		
	90 Days				180 Days			
Sensitivity	-5.578 (0.276)	-5.367 (0.315)	-8.631 (0.114)	3.106 (0.416)	3.899 (0.294)	2.494 (0.529)		
Planned Paydown	0.142** (0.037)	0.151** (0.040)	0.098 (0.219)	0.090 (0.100)	0.115* (0.052)	0.081 (0.252)		
Sensitivity× Sophisticated	-7.314 (0.377)	- <mark>8.092</mark> (0.336)	-5.165 (0.543)	-11.510** (0.023)	-12.603*** -(0.009)	-11.246* -(0.032)		
Planned Paydown× Sophisticated	0.165 (0.219)	0.182 (0.203)	0.244* (0.097)	0.268** (0.024)	0.269** (0.034)	0.320** (0.031)		
Sophisticated	-4.783 (0.265)	-4.912 (0.274)	-7.962* (0.062)	-5.814* (0.064)	-6.105* (0.071)	-8.489*` (0.026)		
Controls								
Median paycheck Original debt Constant	✓ ✓	<b>V V V</b>	<b>\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}}{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}}}}}{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}}}}}}{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac</b>	√ √	<b>\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}}</b>	√ √ √		
Nr of individuals	510	459	408	510	459	408		

P-values of bootstrapped standard errors in parentheses. Significance: (p<0.10), \*\* (p<0.05), \*\*\* (p<0.01).

# ROBUSTNESS: IMPATIENT USERS DRIVE RESULTS



## RELATED LITERATURE

- Theoretical models of present bias
  - Laibson (1997), Laibson, Repetto, Tobacman (2003), O'Donoghue,
     Rabin (1999), Heidhues, Kőszegi (2010)

### Related Literature

- Theoretical models of present bias
  - Laibson (1997), Laibson, Repetto, Tobacman (2003), O'Donoghue,
     Rabin (1999), Heidhues, Kőszegi (2010)
- Empirical evidence of present bias
  - Consumer debt: Meier, Sprenger (2010), Shui, Ausubel (2005), Skiba,
     Tobacman (2008), Ashraf, Karlan, Yin (2006), Bertrand, Morse (2011)
  - Other areas: DellaVigna, Malmendier (2006), Ariely, Wertenbroch (2002), Paserman (2008), Fang, Wang (2010), Choi et al. (2001)

### Related Literature

- Theoretical models of present bias
  - Laibson (1997), Laibson, Repetto, Tobacman (2003), O'Donoghue,
     Rabin (1999), Heidhues, Kőszegi (2010)
- Empirical evidence of present bias
  - Consumer debt: Meier, Sprenger (2010), Shui, Ausubel (2005), Skiba,
     Tobacman (2008), Ashraf, Karlan, Yin (2006), Bertrand, Morse (2011)
  - Other areas: DellaVigna, Malmendier (2006), Ariely, Wertenbroch (2002), Paserman (2008), Fang, Wang (2010), Choi et al. (2001)

This paper: Field data, explicitly measure sophistication

### Related Literature

- Theoretical models of present bias
  - Laibson (1997), Laibson, Repetto, Tobacman (2003), O'Donoghue, Rabin (1999), Heidhues, Kőszegi (2010)
- Empirical evidence of present bias
  - Consumer debt: Meier, Sprenger (2010), Shui, Ausubel (2005), Skiba,
     Tobacman (2008), Ashraf, Karlan, Yin (2006), Bertrand, Morse (2011)
  - Other areas: DellaVigna, Malmendier (2006), Ariely, Wertenbroch (2002), Paserman (2008), Fang, Wang (2010), Choi et al. (2001)

#### This paper: Field data, explicitly measure sophistication

- Sensitivity to paycheck receipt
  - Shapiro (2005), Stephens (2006), Hastings, Washington (2010), Kaur,
     Kremer, Mullainathan (2010), Van Wesep, Parsons (2012)

#### This paper: Sensitivity as proxy for present bias

## DATA

Users of a website to help manage and pay down credit card debt (www.readyforzero.com)

#### DATA

Users of a website to help manage and pay down credit card debt (www.readyforzero.com)

#### Information on:

- Balance, credit limit, available credit of accounts
- Transaction level data for checking and credit card
- Incoming payments, including paycheck
- Amount planned to pay down each month at signup

### DATA

Users of a website to help manage and pay down credit card debt (www.readyforzero.com)

#### Information on:

- Balance, credit limit, available credit of accounts
- Transaction level data for checking and credit card
- Incoming payments, including paycheck
- Amount planned to pay down each month at signup

#### Include users in sample if:

- Linked checking account
- Regular, bi-weekly paychecks
- Data on all linked accounts available from signup till 180 days

## SUMMARY STATISTICS

	N	Mean	p25	p50	p75
Users Days in sample	556	471	339	442	619
Days in sample after sign-up		399	252	354	562
Income					
Avg. monthly income	556	3,913	2,607	3,526	4,669
Assets					
Credit Card Debt - \$	556	15,204	4,962	10,669	19,303
Credit Card Debt - rel. to income		4.52	1.46	3.03	5.10
Cash Balances - \$		3,954	637	1,812	4,452
Total Credit - \$		27,111	9,750	19,875	34,625
Available Credit - \$		11,907	1,776	5,697	16,250
Debt Paydown					
Change in Debt - 90 days - \$	556	-736	-1,332	-234	363
Change in Debt - 90 days - %		0.03	-0.14	-0.02	0.04
Planned Paydown - 90 days - \$		2,747	1,121	1,947	3,484

<sup>►</sup> Spending Summary Statistics