#### Accounting for Households Financial Distress

Kartik Athreya FRB of Richmond Jose Mustre-del-Rio FRB of Kansas City Juan M. Sanchez FRB of St. Louis

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# **Motivation**

• Consider households financial distress defined as:

- 1. Max out available credit card debt.
- 2. Late making credit card payments.
- 3. No wealth.
- Many US household live in financial distress (10-20%).
- Yet, our models miss two key features: life-cycle profile and persistence of financial distress.

## This paper

- Evidence on financial distress (incidence and persistence).
- Show that standard model, calibrated to get the incidence, misses on the persistence.
- Introduce key features to reconcile model and data
  - 1. Persistent expenditure shocks.
  - 2. Permanent discount factor heterogeneity.
  - 3. Informal default.
- Show that these features are important to *get right* the answers to policy relevant questions.

# Life cycle profile of % people with negative net worth



Source: PSID 1998-2010

# Life cycle profile of % of people in default



Source: Equifax 1999-2010

## Life cycle profile of % of people used all credit



Source: Equifax 1999-2010

# Persistence of negative net worth



# Persistence of default



Source: Equifax 1999-2010

# Persistence of "used all credit"



Source: Equifax 1999-2010

### Models' common features

- Incomplete markets and partial equilibrium.
- Households live up to T periods and work until age  $R \le T$ .
- Household's i earnings process has 4 components



where  $z_{i,t} = \rho_z z_{i,t-1} + e_{i,t}$  and the shocks follow Gaussian distributions.

- Post-retirement income depends on the last realization of z.
- · Households cannot commit to repay their debt
- There is a consumption cost (income garnishment) of filing bankruptcy.

# Benchmark model

- Pricing of risk of bankruptcy at the household level.
  - Livshits, MacGee, Tertilt (2007).
  - Chatterjee, Corbae, Nakajima and Rios-Rull (2007) CCNR.
- Additionally:
  - 1. Shocks only to income.
  - 2. Households are ex-ante identical.
  - 3. Only formal default (bankruptcy) allowed prorated like CCNR.

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# Benchmark model vs. Data



# Expense shock model

- Add a stochastic process for health expenditures shocks as estimated by Banks, Blundell, Levell, and Smith (2015).
  - Life-cycle component matches average personal health care expenditures by age from Centers for Medicare & Medicaid Services.
- Recalibrate the model to replicate the (non-prorated) default rate and the share of households in debt.
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#### Health expenditures shocks model vs. Data



# Discount factor heterogeneity model

- Allow for two permanent types in terms of the rate at which households discount the future, β.
- Calibrate the new feature to replicate facts about net worth.
- Recalibrate the model to replicate the default rate, the share (and persistence) of households in debt.

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## Discount factor heterogeneity model vs. Data



# Informal default model

- Allow for two forms of default: formal (bankruptcy) and informal (delinquency).
- In delinquency, households are charged a penalty rate of 20%.
- Bankruptcy involves a period of financial exclusion (exit rate  $\lambda$ ).
- Recalibrate the model to replicate formal and informal default, and the share and persistence of households in debt.

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# Informal default model vs. Data



# Policy implications of alternative models

- Increasing consumption cost of default
- Cap on borrowing rates

# Increasing consumption cost of default by 10 %

	ppt. change in			CE welfare gain (in %)
Model	neg. net worth	default	dq	
Baseline	1.13	-0.15	_	0.06
Expense shocks	1.07	-0.45	_	0.04
Expense + beta-het	1.19	-0.47	_	0.10
Expense + DQBK	0.40	-0.06	-0.09	0.02

### Conclusions

- Standard model cannot account both for the incidence and persistence of financial distress
- Preference heterogeneity and persistent expenditure shocks help reconcile model with data

## Baseline model calibration

Statistic	Target	Model	Parameter	Value
prorated default rate (%)	0.38	0.38	β	0.9735
% in debt	10.82	10.82	τ	55,000



# Expenditure model calibration

Statistic	Target	Model	Parameter	Value
default rate (%)	7.3	7.3	β	0.982
% in debt	10.82	10.82	τ	43,825
Expenditu	ire process	3		
Centers for Medicare & Medicaid Services			φ <sub>n</sub>	0.03
Banks et al 2015			ρχ	0.81
Banks et al 2015			$\sigma_{v}$	0.8

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## Beta-het model calibration

Statistic	Target	Model	Parameter	Value
default rate (%)	7.3	7.3	τ	47,179
% in debt	10.82	12.98	βι	0.89
$Pr(in \ debt_{t+2}  \ in \ debt_t)$	34.97	31.42	$\beta_h$	1.00
$Pr(in \ debt_{t+4}  \ in \ debt_t)$	19.48	20.20	$Pr(\beta_l   \alpha_l)$	0.11
$Pr(in \ debt_{t+6}  \ in \ debt_t)$	12.37	12.37	$Pr(\beta_h   \alpha_h)$	0.69
Expenditu	ire process	5		
Centers for Medicare & Medicaid Services			φn	0.03
Banks et al 2015			ρ <sub>x</sub>	0.81
Banks et al 2015			$\sigma_{\nu}$	0.8



## DQ-BK model calibration

Statistic	Target	Model	Parameter	Value
default rate (%)	0.7	0.7	τ <sub>def</sub>	40,000
delinquency rate (%)	7.3	3.7	τda	0.00093
% in debt	10.82	22.4	β	0.998
$Pr(in \ debt_{t+2}  \ in \ debt_t)$	34.97	18.93	λ	0.93
E	xpenditure proces	s		
Centers for Medicare & Medicaid Ser	vices		φn	0.03
Banks et al 2015			$\rho_X$	0.81
Banks et al 2015			$\sigma_{\nu}$	0.8

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