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Health shock process

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# The Lifetime Costs of Bad Health

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Why bad heal	th is bad?			

## #A. People in bad health

- i. Work less + Earn less if working
- ii. Face higher medical expenses
- iii. Have lower life expectancy

## #B. Over the life cycle, the *accumulated* effects of bad health

- Depend on how long the sickness lasts
- Can be substantial when health is persistent and markets are incomplete

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Is the *accumulated* effect important?

Wealth-health gradient among high school men (HRS: 1994-2012)



- good health  $\in$  {*excellent*, *very good*, *good*}; bad health  $\in$  {*fair*, *poor*}

- net worth: controlled for year effects and family sizes

The difference is large even among a relatively homogeneous group

wealth change

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Our stand on health and economic outcomes

How do economists think about health and economic outcomes?

Ch.1 Health is *exogenous*: health  $\Rightarrow$  economic outcomes

Ch.2 Health is *endogenous*: economic outcomes  $\Rightarrow$  health

Ch.3 People differ in factors affecting both their health and economic outcomes

childhood circumstances

genetics

This paper

 $\Rightarrow$  focus on Ch.1 and 3

⇒ quantify effects of health uncertainty under incomplete markets

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What we do?	The big picture			

1<sup>st</sup> **Part** : Why is health status persistent?

- Document long-term dynamics of health status in the data
- Estimate a parsimonious health shock process that is consistent with the empirical facts (both cross-sectional and dynamic aspects)
- Identify two different sources of health persistence
  - i. Duration-dependence: the longer an unhealthy spell, the lower the chance of recovering
  - ii. Fixed health type: people are different, eg. lifestyle, genes

2<sup>nd</sup> Part: How does bad health affect individuals over life cycle?

- Estimate a life cycle model augmented with the health shock that captures
  - 1. Effects of bad health on life expectancy and medical spending
  - 2. Income-health gradient
  - 3. Wealth-health gradient
- And answer the following questions
  - i. How much is the monetary loss due to bad health over life cycle?
  - ii. Why being in good health is valuable?
  - iii. How much does health uncertainty contribute to lifetime inequality?

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Data				

- 1. Health and Retirement Study (HRS: 1994-2012)
- 2. Panel Study of Income Dynamics (PSID)
  - Annual data (1984-1997); bi-annual (1997-2012)
- 3. Medical Expenditure Panel Survel (MEPS: 1999-2011)

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Outline of the	presentation			



► Life-cycle model

► Model estimation (MSM)





20-24 25-29 30-34 35-39 40-44 45-49 50-54 55-59 60-64 65-69 70-74 75+ ages





#### <u>Panel B2</u>: % transition $good \rightarrow bad$





### Duration-dependent profile by health status (30-54 years old)



Long duration-dependence



Panel C2: % Transition from good to bad health





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Health sho	ock process			

Conditional on surviving to the next period,

Probability to be *healthy* if unhealthy for  $\tau_B$  yrs:  $\pi_i^{\vec{BG}}(\tau_B|age)$ 

$$logit\left(\pi_{i}^{\overrightarrow{BG}}(\tau_{B}|age)\right) = \underbrace{\left(a_{1}^{B}\mathbf{1}_{\{\tau_{B}=1\}} + a_{2}^{B}\mathbf{1}_{\{\tau_{B}\geq2\}}\right)}_{\text{duration dependence}} + \underbrace{\left(b_{1}^{B}age + b_{2}^{B}age^{2}\right)}_{\text{health type}} + \underbrace{\eta_{i}}_{\text{health type}}$$

Probability to be *unhealthy* if healthy for  $\tau_G$  yrs:  $\pi_i^{\vec{GB}}(\tau_G|_{age})$ 

$$\textit{logit}\left(\pi_i^{\overrightarrow{GB}}(\tau_G|\textit{age})\right) = \left(a_1^G \mathbf{1}_{\{\tau_G=1\}} + a_2^G \mathbf{1}_{\{\tau_G \ge 2\}}\right) + \left(b_1^G \textit{age} + b_2^G \textit{age}^2\right) + b_3^G \times \eta_i$$

 $\eta_i \sim$  uniform distribution over 5 points symmetric around zero

surv prob



>=1vr



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#### Estimated health shock process

#### *bad*⇒*good*



 $\rightarrow$  Most of duration dependence is due to fixed health type

#### $good \Rightarrow bad$



 $\rightarrow$  No effect of fixed health type

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#### Distribution of unhealthy periods between 57-65: Model vs HRS

#### (Additional validation)



HRS: balanced panel of healthy individuals at 55 (N=828 individuals)

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How should we	think about health	type?		

Model: People with bad health type experience multiple periods being unhealthy

► *HRS:* Characteristics of people by #periods being unhealthy

# unhealthy yrs	$\% \eta_1 + \eta_2$	% smoking	BMI <sup>a</sup>	% pare	ent alive	parents'	educ (yrs)	PGS
(57-65)	(model)			father	mother	father	mother	Educ <sup>b</sup>
0-1	26.9	23.2	27	21.2	49.5	10	12	-0.10
2-3	39.7	25.9	28	20.2	46.7	9	10	-0.18
4-5	71.1	43.5	30	15.2	36.9	8	8	-0.64

Individuals are healthy at 55

<sup>a</sup> BMI=body mass index (median)

<sup>b</sup> PolyGenetic Score for Educational Attainment

 $\rightarrow$  labor market outcomes (Papageorge and Thom, 2019)

 $\rightarrow$  genetic-wealth gradient (Barth, Papageorge and Thom, 2019)

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## Life-cycle model

► Model estimation (MSM)



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Key mechanism	าร			

- The observed correlation between health and life-cycle outcomes is generated by two mechanisms
- $1\,$  Causal effects of bad health:
  - a. Decreases productivity and increases disutility from work
  - b. Increases OOP medical spending
  - c. Lowers life expectancy
- 2 Composition effect:
  - Fixed and heterogenous health types  $(\eta_i)$
  - Fixed and heterogenous patience  $(\beta_i)$
  - $\eta_i$  and  $\beta_i$  can be correlated.

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Life-cycle mod	el			

▶ 20-64 $\rightarrow$ work, 65-99 $\rightarrow$ retired

▶ health type:  $\eta_i \in \{\eta_1, ..., \eta_5\}$  and discount factor:  $\beta_i \in \{\beta_{low}, \beta_{high}\}$  $0 \le Pr(\beta_i | \eta_m) \le 1; j \in \{low, high\}, m \in \{1, 2, ..., 5\}$ 

- People face productivity, health, medical expenses, and survival uncertainty
- Retired people receive Social Security benefits and are covered by Medicare



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Health process estimation

## Life-cycle model

- Model estimation (MSM)
  - wealth profile
  - employment profile + average labor income profile



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## Model parameters taken/estimated outside model

parameters		sources
Survival probability by health:	$\zeta_t^h$	HRS
		(extrapolation from 20 to 50)
Health transition probability:	$\pi_{i,t}^{\overrightarrow{BG}}\left(\tau_{B}\right), \ \pi_{i,t}^{\overrightarrow{GB}}\left(\tau_{G}\right)$	PSID
Labor productivity shock:	$z_{i,t}^h$	PSID
Health-dependent medical expenses:	x <sub>t</sub> <sup>h</sup>	MEPS
ESI offer probability (logit) :	$g_t^{h,z}$	MEPS
Insurance coverage:	$cvg(x_t^h, i_H)$	MEPS
Risk aversion:	ho = 3.0	common values $\in [1, 5]$

#### Parameters taken/estimated outside model



Stochastic processes estimated outside the model

• Health-dependent labor income process  $(z_t^h)$ 

$$\begin{aligned} z_{i,t}^{h} &= \lambda_{t}^{h} + \gamma_{i} + y_{i,t} \\ y_{i,t} &= \rho_{y} y_{i,t-1} + \varepsilon_{i,t}; \quad \varepsilon_{i,t} \sim \textit{iid } N\left(0, \sigma_{\varepsilon}^{2}\right) \end{aligned}$$

From PSID:  $\rho_y = 0.9275, \ \sigma_{\varepsilon}^2 = 0.0209, \ \sigma_{\gamma}^2 = 0.042$ 

 λ<sup>h</sup><sub>t</sub> is used to match average labor income among healthy and unhealthy workers

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Parameters e	estimated inside mo	odel		

parameters value			targets			
$\{\beta_{low}, \beta_{high}\}$	$\{0.904, 0.995\}$				"	
$Pr(\beta_{low} n_i)$	$\eta_1$	$\eta_2$	$\eta_3$	$\eta_4$	$\eta_5$	net wealth profiles
(, , , , , , , , , , , , , , , , , , ,	0.89	0.81	0.66	0.36	0.12	by health (PSID)
consumption floor: $\overline{c}$ \$3,593 (or \$5,484 in 2010)			"			

\*  $\eta_1$  has the lowest probability to recover

- $\overline{b} \Rightarrow$  Statistical Value of Life (SVL)
  - Compensation for adding 1 death among 10,000 adults:
  - Empirical SVL = 1-16M USD
  - Model: average SVL among working-age individuals = 2M USD



#### at median



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## The importance of compositional difference

Wealth difference between healthy and unhealthy people at ages 60-64.

Wealth difference by health	PSID	Baseline	No $(\beta_{low}, \eta_i)$ correlation
25 <sup>th</sup> pct	41,225	54,157	32,497
50 <sup>th</sup> pct	97,142	101,094	39,715
75 <sup>th</sup> pct	156,824	146,225	70,404

 No correlation between types and patience misses health-wealth gradient

- Income-health gradient does not imply wealth-health gradient

▶ details

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Results				

- R1. The monetary cost of bad health during the working period
- R2. The value of being in good health 📀
- R3. The contribution of health to lifetime inequality

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R1. The m	onetary cost of bad	health		

## Exp#1:

- Everyone always draws good health
- Consider those surviving to age 64 in baseline
- Monetary costs<sub>it</sub> of bad health =

earnings loss<sub>it</sub> +medical costs<sub>it</sub> (during 20 to 64)

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#### R1. The monetary cost of bad health

Average loss (per year) over 20-64



avg labor income=\$36,105

- Varies a lot by health type
- Health insurance covers a non-trivial portion of the cost
- Earning loss is much larger than OOP medical loss

by unhealthy years

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#### R1. The monetary loss due to bad health

## Distribution of lifetime cost of bad health

	% of total		
	top 5%	top 10%	top 20%
earning loss + total medical loss	28%	46 %	71%
earning loss $+$ OOP loss	27%	45 %	72%

#### Highly concentrated

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 R2. The value of being in good health (20-64)

Exp#2:

- Increase the probability of being in good health by 1% from period t to t + 1
- Calculate willingness to pay to move from the baseline to the experiment above (among people aged 20-64)

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## R2. The value of being in good health (20-64)

## Sources of the gains

- 1. Allow one channel through which health affects individuals
- 2. Recompute the remaining gain

	$\eta_1 - \eta_5$	$\eta_1$	$\eta_3$	$\eta_5$		
Baseline economy	\$1,903	\$2,933	\$1,718	\$1,200		
(% of avg labor inc)	(5.3%)	(8.1%)	(4.8%)	(3.3%)		
Dollar value when only one chann	Dollar value when only one channel exists					
- Survival channel	60%	52%	61%	74%		
- Labor market channel	36%	45%	34%	22%		
- Medical expenses channel	5%	5%	4%	4%		

% is a fraction of willingness to pay in the baseline

## Survival channel contributes most to the value of being healthy





- Everyone always draws good health till death
  - Case 1. Allow age of death to increase  $\Rightarrow$  include survival channel
  - Case 2. Fix age of death as in Baseline  $\Rightarrow$  exclude survival channel
- Define Lifetime utility

$$U_i = \sum_{t=20}^{\text{age of death}+1} \beta_i^{t-20} \Big( u(c_t, l_t, h_t) \times 1_{\text{alive}_t} + Beq_t \times (1 - 1_{\text{alive}_t}) \Big)$$

► Variation of  $U_i$  due to health  $= \left(1 - \frac{V(\hat{U}_i)}{V(U_i^B)}\right) \times 100\%$ 

$$\hat{U}_i =$$
 lifetime utility from R3  
 $U_i^B =$  lifetime utility from Baseline

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## R3. Lifetime inequality due to health

Case 1. Include survival channels (allowing age of death to increase)

	$\beta_{low}$	$eta_{high}$
all $\eta_i$	47%	14%
$\Rightarrow \{\eta_1, \eta_2\}$	54%	25%
$\Rightarrow \{\eta_3, \eta_4, \eta_5\}$	30%	10%

Variation of lifetime utility due to health

Case 2. Exclude survival channels (fixing age of death as in Baseline) Variation of lifetime utility due to health

	$\beta_{low}$	$eta_{high}$
all $\eta_i$	24%	4%
$\Rightarrow \{\eta_1, \eta_2\}$	28%	11%
$\Rightarrow \{\eta_3, \eta_4, \eta_5\}$	11%	1%

\*  $\eta_1, \eta_2$  have lower probability to recover

- Survival channel attributes a lot to lifetime inequality
- Health affects lifetime ineq. more among those with bad health type (η<sub>1</sub>, η<sub>2</sub>)

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#### Conclusions

- We quantify the effects of health in a life-cycle model of high school males that matches
  - (1) Long-run health dynamics
  - (2) Income-health gradient
  - (3) Wealth-health gradient
- Health type: important for capture (1)
- Compositional difference btw. the healthy and unhealthy: important for (3)
- Implications
  - i. Lifetime costs of bad health are highly concentrated
  - ii. The earning losses due to bad health are the largest component of OOP losses
  - iii. The most valuable aspect of being healthy is a longer life expectancy
  - iv Survival channel attributes a lot to lifetime inequality

#### Distribution of unhealthy periods between 57-65 (HRS)



HRS: balanced panel of healthy individuals at 55 (N=828 individuals)

## A non-trivial fraction experiences multiple periods being unhealthy

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### Dynamic wealth-health gradient (HRS)



The longer being unhealthy, the lower accumulation of wealth

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## Health-dependent survival probability



Estimated health-dependent survival probability (HRS: 1994-2012)

## Sample from PSID: 1984-1997

% Transition from bad to good health conditioned on being in bad health						
	>= 1	>= 2	>= 3	>= 4	>= 5	>= 6
		num	ber of indivi	idual-years		
30-54	1106	602	389	271	201	149
55-69	568	364	253	180	129	92
70+	429	247	156	101	69	46
number of individuals						
30-54	376	196	123	79	60	43
55-69	163	106	73	53	38	28
70+	125	78	51	32	23	17

#### % Transition from good to bad health conditioned on being in good health

	>= 1	>= 2	>= 3	>= 4	>= 5	>= 6
		num	ber of indivi	dual-years		
30-54	8089	6668	5524	4578	3789	3115
55-69	1791	1452	1205	1008	843	697
70+	734	515	376	281	210	156
		п	imber of ind	lividuals		
30-54	1267	1125	987	847	735	666
55-69	326	268	222	191	169	157
70+	160	118	89	68	54	42

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### Dynamics of health status data (PSID vs PSID excl. DI)

Duration-dependent profile by health status (30-54 years old, excl. DI)



▶ back

### Dynamics of health status (PSID vs PSID excl. DI)



## Model: working-age individuals

## Consumption-saving problem

$$\max_{c_t,k_{t+1}} u(c_t,l_t,h_t) + \beta_i \left( \zeta_t^{\ h} E_t V_{t+1}^i(\mathbb{S}_{t+1}) + \left(1 - \zeta_t^{\ h}\right) \theta_{Beq} \left(\frac{k_{t+1} + k_{Beq}}{1 - \rho}\right)^{1 - \rho} \right)$$

$$\underbrace{k_t (1+r)}_{\text{total asset}} + \underbrace{exp \left(z_{it}^h\right) \ l_t}_{\text{labor inc}} - \text{OOP med}_{it} - \text{Ins prem} - Tax + T^{SI}(\overline{c}) = c_t + k_{t+1}$$

back

Health-dependent total medical expenses  $(x_t^h)$ 

 $\blacktriangleright$   $x_t^h$  is directly estimated from MEPS



cvg(x<sub>t</sub><sup>h</sup>, i<sub>H</sub>) is estimated from people with ESI or ind insurance
 g<sub>t</sub><sup>h,z</sup> is parameterized as a logit function and estimated from MEPS

## Targeted moments: Model vs PSID

#### Health and labor market outcomes



% Workers by health status

Average labor income (among workers) by health



## Implied health gradients: Model vs PSID (HRS)

		PSID (HRS)			Model	
	bottom 1/3	middle 1/3	top 1/3	bottom 1/3	middle 1/3	top 1/3
25-34	12%	5%	2%	16%	2%	0%
35-44	21%	8%	4%	22%	4%	2%
45-54	22%	12%	8%	28%	9%	5%
55-64	30% (36%)	15% (20%)	14% (13%)	33%	24%	11%

## % unhealthy individuals in each earnings tercile

% unhealthy individuals in each wealth tercile

		PSID (HRS)			Model	
	bottom 1/3	middle 1/3	top 1/3	bottom 1/3	middle 1/3	top 1/3
25-34	10%	10%	5%	8%	5%	3%
35-44	17%	10%	5%	14%	7%	5%
45-54	23%	13%	9%	24%	10%	8%
55-64	33% (36%)	17% (21%)	12% (14%)	34%	17%	13%
65-74	36% (38%)	26% (24%)	17% (16%)	41%	27%	19%
75+	46% (41%)	37% (29%)	24% (25%)	47%	38%	29%

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### Implied dynamic wealth-health gradient: Model vs HRS

Median wealth change between 55/56 and 65/66



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#### Implied dynamic wealth-health gradient: Model vs HRS



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## The importance of types - health and patience

Wealth-health gradient (60-64)					
Wealth difference	PSID (HPS)	Bacolino	No correlation		
by health	1312 (11(3)	Dasenne	$Pr\left(eta_{low} \eta_{i} ight)=0.5$		
25 <sup>th</sup> pct	41,225 (47,569)	54,157	32,497		
50 <sup>th</sup> pct	97,142 (92,726)	101,094	39,715		
75 <sup>th</sup> pct	156,824 (178,466)	146,225	70,404		

(Unconditional)	) wealth dist (60-64)	)
	<b>D</b>	No

Wealth level	PSID (HRS)	Baseline	$Pr\left(\beta_{low} \eta_i\right) = 0.5$
25 <sup>th</sup> pct	75,997 (76,253)	83,041	86,652
50 <sup>th</sup> pct	169,557 (165,454)	180,525	187,746
75 <sup>th</sup> pct	343,298 (349,858)	339,387	346,608
$\beta_i$	-	{0.90, 0.99}	{0.90, 0.99}
ī	-	\$3593	\$3540
$\theta_{Beq}, k_{Beq}$	-	${4464, 246371}$	<b>{4370</b> , 228476 <b>}</b>

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#### R1. The monetary cost of bad health

Average loss (per year) over 20-64



avg labor income=\$36,105

#### Increases steeply with the number of unhealthy years

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## R2. The value of being in good health by asset terciles (20-64)

	1 <sup>st</sup> Tercile	Asset terciles 2 <sup>nd</sup> Tercile	3 <sup>rd</sup> Tercile	
Baseline economy (% avg labor income)	\$1,333 <i>(3.7%)</i>	\$1,770 <i>(4.9%)</i>	\$2,453 <i>(6.8%)</i>	
Dollar value when only one channel exists				
- Survival channel	35%	47%	78%	
- Labor market channel	58%	45%	21%	
- Medical expenses channel	7%	5%	3%	

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## R2. The value of being in good health (20-64) when SVL =\$6M

	$\eta_1 - \eta_5$	$\eta_1$	$\eta_3$	$\eta_5$
Baseline economy (% of avg labor inc)	\$3,828 <i>(10.6%)</i>	\$5,113 (14.1%)	\$3,506 <i>(9.7%)</i>	\$3,026 <i>(8.4%)</i>
Dollar value when only one channel exists				
- Survival channel - Labor market channel - Medical expenses channel	86% 18% 2%	81% 26% 3%	86% 16% 2%	93% 9% 1%

% is a fraction of willingness to pay in the first row

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### R3. Lifetime inequality due to health *when SVL=\$6M*

Case 1. Exclude survival channels (fixing age of death as in Baseline)

valiation of meetine attinty due to nearth		
	$\beta_{low}$	$eta_{high}$
all $\eta_i$	7.35%	0.22%
$\Rightarrow \{\eta_1, \eta_2\}$	9.5%	0.7%
$\Rightarrow \{\eta_3, \eta_4, \eta_5\}$	2.6%	0.0%

Variation of lifetime utility due to health

\*  $\eta_1, \eta_2$  have lower probability to recover

Case 2. Include survival channels (allowing age of death to increase)

	$\beta_{low}$	$eta_{high}$
all $\eta_i$	42.5%	12.8%
$\Rightarrow \{\eta_1, \eta_2\}$	47.5%	20.2%
$\Rightarrow \{\eta_3, \eta_4, \eta_5\}$	33.3%	9.9%

#### Variation of lifetime utility due to health