

# Health, Consumption, and Inequality

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Work in Progress (still)

## Motivation

- Inequality is one of the themes of our time.
  - Large body of literature documenting inequality in **labor earnings**, **income**, and **wealth** across countries and over time  
Katz, Murphy (QJE 1992); Krueger *et al* (RED 2010); Piketty (2014)  
Kuhn, Ríos-Rull (QR 2016); Khun *et al* (2017)
- We also know of large **socio-economic gradients in health outcomes**
  - In mortality  
Kitagawa, Hauser (1973); Pijoan-Mas, Rios-Rull (Demography 2014)  
De Nardi *et al* (ARE 2016); Chetty *et al* (JAMA 2016)
  - In many other health outcomes  
Marmot *et al* (L 1991); Smith (JEP 1999)  
Bohacek, Bueren, Crespo, Mira, Pijoan-Mas (2018)
- ▷ We want to *compare* and *relate* **inequality in health outcomes** to pure **economic inequality**.

## The project

- 1 Write a **model** of **consumption**, **saving** and **health choices** featuring
  - (a) Health-related preferences
  - (b) Health technology
- 2 Use the FOC (only) to estimate (a) and (b)
  - **Consumption growth** data to estimate how health affects the marginal utility of consumption
  - Standard measures of **VSL** and **HRQL** to infer how much value individuals place on their life in different health states
  - Medical **health spending**, **health transitions** (and people's valuation of life) to infer health technology
- 3 Use our estimates to
  - Welfare analysis: compare different groups given their allocations
  - Ask what different groups would do if their resources were different and how much does welfare change
  - Evaluate public policies?

## Main empirical challenge

- Theory:
  - Out-of-pocket expenditures improve health
- Data:
  - Cross-section: higher spending leads to better health transitions across groups (education, wealth)
  - Panel: higher spending leads to worse outcomes
    - ▷ unobserved health shocks spur medical spending
- Add explicitly into the model
  - Unobserved shock to health between  $t$  and  $t + 1$  that shapes
    - probability of health outcomes
    - the returns to health spending
  - Higher expenditure signals higher likelihood of bad health shock

Model

## Life-Cycle Model (mostly old-age)

- 1 Individuals state  $\omega \in \Omega \equiv I \times E \times A \times H$  is
  - Age  $i \in I \equiv \{50, \dots, 89\}$
  - Education  $e \in E \equiv \{\text{HSD}, \text{HSG}, \text{CG}\}$
  - Net wealth  $a \in A \equiv [0, \infty)$
  - Overall health condition  $h \in H \equiv \{h_g, h_b\}$
  
- 2 Choices:
  - Consumption  $c \in \mathbb{R}_{++} \rightarrow$  gives utility
  - Medical spending  $x \in \mathbb{R}_+ \rightarrow$  affects health transitions
  - Next period wealth  $a' \in A$
  
- 3 Shocks:
  - Unobserved health outlook shock  $\eta$
  - Implementation error  $\epsilon$  in health spending
  
- 4 (Stochastic) Health technology:
  - Health transitions given by  $\Gamma^{ei}[h' \mid h, \eta, x\epsilon]$
  - Survival given by  $\gamma^i(h)$  (note no education or wealth)

## Uncertainty and timing of decisions

- ① At beginning of period  $t$  individual state is  $\omega = (i, e, a, h)$
- ② Consumption  $c$  choice is made
- ③ *Health outlook shock*  $\eta \in \{\eta_1, \eta_2\}$  with probability  $\pi_\eta$
- ④ Health spending decision  $x(\omega, \eta)$  is made
- ⑤ *Medical treatment implementation shock*  $\log \epsilon \sim N\left(-\frac{1}{2}\sigma_\epsilon^2, \sigma_\epsilon^2\right)$ 
  - Once health spending is made, the shock determines actual treatment obtained  $\tilde{x} = x(\omega, \eta) \epsilon$
  - Allows for the implementation of the Bayesian updating of who gets the bad health outlook shock

## The Bellman equation

### The retiree version

- The household chooses  $c$ ,  $x(\eta)$ ,  $y(\eta)$  such that

$$v^{ei}(h, a) = \max_{c, x(\eta), y(\eta)} \left\{ u^i(c, h) + \beta^e \gamma^i(h) \sum_{h', \eta} \pi_{\eta}^{ih} \int_{\epsilon} \Gamma^{ei}[h' | h, \eta, x(\eta)\epsilon] v^{e, i+1}[h', a'(\eta, \epsilon)] f(d\epsilon) \right\}$$

- s.t. the budget constraint and the law of motion for cash-in-hand

$$\begin{aligned} c + x(\eta) + y(\eta) &= a \\ a'(\eta, \epsilon) &= [y(\eta) - (\epsilon - 1)x(\eta)]R + w^e \end{aligned}$$

- The FOC give:
  - One Euler equation for consumption  $c$
  - One Euler equation for health investments at each state  $\eta$



## FOC for consumption

- Optimal choice of consumption for individuals of type  $\omega$
- Standard Euler equation for consumption w/ sophisticated expectation (Over survival, health tomorrow  $h'$ , outlook shock  $\eta$ , and implementation shock  $\epsilon$ )

$$u_c^i[h, c(\omega)] = \beta^e \gamma^i(h) R \sum_{h', \eta} \pi_{\eta}^{ih} \int_{\epsilon} \Gamma^{ei}[h' | h, \eta, x(\omega, \eta)\epsilon] u_c^{i+1}[h', c(\omega, \eta, h', \epsilon)] f(d\epsilon)$$

- Timing assumptions  $\Rightarrow$  consumption independent from shocks  $\eta, \epsilon$
- Then, it is easy to estimate w/o other parts of the model:
  - *expected transitions are the same for all individuals of same type  $\omega$*

## FOC for health spending

- Individuals of type  $\omega$  make different health spending choices  $x(\omega, \eta)$  depending on their realized  $\eta$
- The FOC for individual of type  $\omega$  is  $\eta$ -specific:

$$\sum_{h'} \int_{\epsilon} \underbrace{\epsilon \Gamma_x^{ei}[h' | h, \eta, x(\omega, \eta)\epsilon]}_{\text{improvement in health transition}} \underbrace{v^{e,i+1}\{h', a'(\omega, \eta, \epsilon)\}}_{\text{value of life tomorrow}} f(d\epsilon) =$$

$$\underbrace{R \sum_{h'} \int_{\epsilon} \epsilon \Gamma_c^{ei}[h' | h, \eta, x(\omega, \eta)\epsilon] u_c^{i+1}[h', c(\omega, \eta, h', \epsilon)] f(d\epsilon)}_{\text{Expected utility cost of forgone consumption}}$$

- In order to use this for estimation we need to
  - Allocate individuals to some realization for  $\eta$
  - Compute the value function

# Estimation

## Preliminaries

- We group wealth data  $a_j$  into quintiles  $p_j \in P \equiv \{p_1, \dots, p_5\}$ 
  - State space is the countable set  $\widehat{\Omega} \equiv E \times I \times H \times P$

- Functional forms

- Utility function

$$u^i(h, c) = \alpha_h + \chi_h^i \frac{c^{1-\sigma_c}}{1-\sigma_c}$$

- Health transitions

$$\Gamma^{ie}(g|h, \eta, x) = \lambda_{0\eta}^{ieh} + \lambda_{1\eta}^h \frac{x^{1-\nu^h}}{1-\nu^h}$$

- Estimate several transitions in HRS data

- Survival rates  $\tilde{\gamma}_h^i$
- Health transitions  $\tilde{\Gamma}(h_g|\omega)$
- Health transitions conditional on health spending  $\tilde{\varphi}(h_g|\omega, \tilde{x})$
- Joint health and wealth transitions  $\tilde{\Gamma}(h', p'|\omega)$

## General strategy

- Estimate vector of parameters  $\theta$  by GMM without solving the model
  - Use the restrictions imposed by the FOC
  - Need to compute value functions with observed choices and transitions

- Two types of parameters

1/ Preferences:  $\theta_1 = \{\beta^e, \sigma_c, \chi_h^i, \alpha_h\}$

- Can be estimated independently from other parameters
- Use **consumption Euler equation** to obtain  $\beta^e, \sigma_c, \chi_h^i$
- Use **VSL and HRQL conditions** to estimate  $\alpha_h$

2/ Health technology:  $\theta_2 = \{\lambda_{0\eta}^{ieh}, \lambda_{1\eta}^h, \nu^h, \pi_\eta, \sigma_\epsilon^2\}$

- Requires  $\theta_1$  as input
- Use **medical spending Euler equations plus health transitions**
- Problem: we observe neither  $\eta_j$  nor  $\epsilon_j$
- Need to recover posterior probability of  $\eta_j$  from observed health spending  $\tilde{x}_j$

## Data: various sources

### 1 HRS

- White males aged 50-88
- Health stock measured by **self-rated health** (2 states)
- ▷ Obtain the objects  $\tilde{\gamma}_h^i$ ,  $\tilde{\Gamma}(h_g|\omega)$ ,  $\tilde{\varphi}(h_g|\omega, \tilde{x})$ ,  $\tilde{\Gamma}(h', p'|\omega)$

### 2 PSID (1999+) gives

- Households headed by white males aged 50-88
- Non-durable consumption
- Out of Pocket medical expenditures

### 3 Standard data in clinical analysis

- Outside estimates of the value of a statistical life (VSL)
- Health Related Quality of Life (HRQL) scoring data from HRS

## Preliminary Estimates: Preferences

## Marginal utility of consumption

### Consumption Euler equation

- We use the sample average for all individuals  $j$  of the same type  $\omega$  as a proxy for the expectation over  $\eta$ ,  $h'$ , and  $\epsilon$

$$\beta^e R \tilde{\gamma}_h^i \frac{1}{N_\omega} \sum_j \mathbf{1}_{\omega_j=\omega} \frac{\chi_{h_j}^{i+1}}{\chi_h^i} \left( \frac{c_j'}{c_j} \right)^{-\sigma} = 1 \quad \forall \omega \in \tilde{\Omega}$$

- Normalize  $\chi_g^i = 1$  and parameterize  $\chi_b^i = \chi_b^0 (1 + \chi_b^1)^{(i-50)}$
  - Use cons growth from PSID by educ, health, wealth quintiles
- We obtain
  - Health and consumption are complements  
Finkelstein et al (JEEA 2012), Koijen et al (JF 2016)
  - More so for older people
  - Uneducated are NOT more impatient: they have worse health outlook



# Marginal utility of consumption

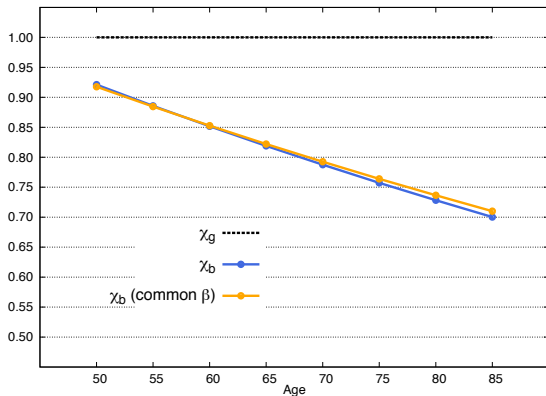
## Results

Men sample (with $r = 4.04\%$ )				
	$\beta$ edu specific		$\beta$ common	
$\sigma$	1.5		1.5	
$\beta^d$ (s.e.)	0.8861	(0.0175)	0.8720	(0.0064)
$\beta^h$ (s.e.)	0.8755	(0.0092)	0.8720	(0.0064)
$\beta^c$ (s.e.)	0.8634	(0.0100)	0.8720	(0.0064)
$\chi_b^0$ (s.e.)	0.9211	(0.0575)	0.9176	(0.0570)
$\chi_b^1$ (s.e.)	-0.0078	(0.0035)	-0.0073	(0.0035)
observations	15,432		15,432	
moment conditions	240		240	
parameters	5		3	

Notes: estimation with biennial data. Annual interest rate of 2%, annual  $\beta$ : 0.9413, 0.9357, 0.9292 in first column and 0.9338 in the second one.

# Marginal utility of consumption

## Results



$$\frac{C_g}{C_b} = \left( \frac{\chi_g}{\chi_b} \right)^{1/\sigma} = \begin{cases} 1.057 & \text{at age 50} \\ 1.268 & \text{at age 85} \end{cases}$$

## Value of life in good and bad health

We use standard measures in clinical analysis to obtain  $\alpha_g$  and  $\alpha_b$

### ① Value of Statistical Life (VSL)

- From wage compensation of risky jobs [Viscusi, Aldy \(2003\)](#)
- Range of numbers: \$4.0M–\$7.5M to save one statistical life
- This translates into \$100,000 per year of life saved
- ▷ Calibrate the model to deliver same MRS between survival probability & cons flow [Becker, Philipson, Soares \(AER 2005\)](#); [Jones, Klenow \(AER 2016\)](#)

### ② Quality Adjusted Life Years (QALY)

- Trade-off between years of life under different health conditions
- From patient/individual/household surveys: *no revealed preference*
- Use HUI3 data from a subsample of 1,156 respondents in 2000 HRS
- Average score for  $h = h_g$  is 0.85 and for  $h = h_b$  is 0.60
- ▷ Calibrate the model to deliver same relative valuation of period utilities in good and bad health

## The value functions

- The value achieved by an individual of type  $\omega$  is given by

$$v^{ei}(h, a) = u^i(c(\omega), h) + \beta^e \gamma^i(h) \sum_{h', \eta} \pi_{\eta}^{ih} \int_{\epsilon} \Gamma^{ei}[h'|h, \eta, x(\omega, \eta), \epsilon] v^{ei+1}(h', a'(\omega, \eta, \epsilon)) f^x(d\epsilon)$$

with

$$a'(\omega, \eta, \epsilon) = (a - c(\omega) - \epsilon x(\omega, \eta))(1 + r) + w^e$$

- We can compute the value function from observed choices and transitions *without solving for the whole model* by rewriting the value function in terms of wealth percentiles  $p \in P$ :

$$v^{ei}(h, p) = \frac{1}{N_{\omega}} \sum_j \mathbf{1}_{\omega_j = \omega} u^i(c_j, h_j) + \beta^e \tilde{\gamma}_h^i \sum_{h', p'} \tilde{\Gamma}[h', p' | \omega] v^{ei+1}(h', p')$$

where we have replaced the expectation over  $\eta$  and  $\epsilon$  by the joint transition probability of assets and health,  $\tilde{\Gamma}[h', p' | \omega]$

Preliminary Estimates: health technology

## The moment conditions: Preview

- For each  $\omega = (i, e, h, p)$ , we have four distinct moment conditions.
  - (M1) Health spending EE for  $\eta_g$
  - (M2) Health spending EE for  $\eta_b$
  - (M3) Average Health transitions for  $x > \text{median}(x_\omega)$
  - (M4) Average Health transitions for  $x < \text{median}(x_\omega)$
  
- We have  $210 \times 4 = 840$  moment conditions
  - $e$ : 3 edu groups = {HSD, HSG, CG}
  - $i$ : 8 age groups = {50-54, 55-59, 60-64, 65-69, 70-74, 75-79, 80-84, 85-89}
  - $h$ : 2 health groups =  $\{h_g, h_b\}$
  - $p$ : 5 wealth groups
  - ▷ This gives 240 cells in  $\omega$
  - But there are 30 cells that are empty (20 in age 85+, 5 in age 80-84)

## The Problem

- Key problem: how to deal with unobserved health shock  $\eta$ 
  - Needed to evaluate the moment conditions (M1) to (M4)
- We construct the **posterior probability of  $\eta$**  given observed health investment  $\tilde{x}_j$  and the individual state  $\omega_j$

$$Pr[\eta_g | \omega_j, \tilde{x}_j] = \frac{Pr[\tilde{x}_j | \omega_j, \eta_g] Pr[\eta_g | \omega_j]}{Pr[\tilde{x}_j | \omega_j]}$$

- where  $Pr[\tilde{x}_j | \omega_j, \eta_g]$  is the density of  $\epsilon_j = \tilde{x}_j/x(\omega_j, \eta_g)$
- where  $Pr[\eta_g | \omega_j] = \pi_{\eta_g}$
- where  $Pr[\tilde{x}_j | \omega_j] = \sum_{\eta} Pr[\tilde{x}_j | \omega_j, \eta] Pr[\eta | \omega_j]$
- We weight every individual observation by this probability

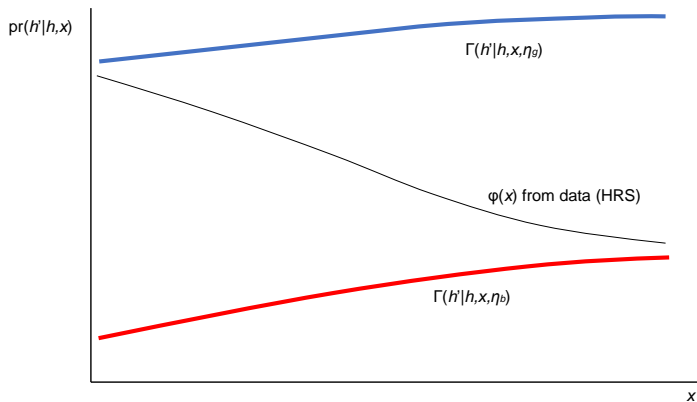
## The Problem

- To obtain the posterior distributions we need to estimate
  - the contingent health spending rule,  $x(\omega, \eta)$
  - the variance of the medical implementation error,  $\sigma_\epsilon^2$
  - the probability distribution of health outlooks sock,  $\pi_{\eta_g}$
- We identify all these objects through the observed health transitions  $\tilde{\varphi}(h_g|\omega, \tilde{x})$  as function of the state  $\omega$  and health spending  $\tilde{x}$

$$\underbrace{Pr[h_g|\omega, \tilde{x}]}_{\text{observed in the data}} = \Gamma[h_g | \omega, \eta_g, \tilde{x}] \underbrace{Pr[\eta_g|\omega, \tilde{x}]}_{\text{posterior}} + \Gamma[h_g | \omega, \eta_b, \tilde{x}] \underbrace{(1 - Pr[\eta_g|\omega, \tilde{x}])}_{\text{posterior}}$$



# The Problem



## Moment conditions

### Health Spending Euler Equation

- Moment conditions (M1) to (M2) identify the curvature  $\nu^h$  and slope  $\lambda_{1\eta}^h$  of the health technology
- $\forall \omega \in \tilde{\Omega}$  and  $\forall \eta \in \{\eta_g, \eta_b\}$  we have

$$\frac{1}{M_{\omega\eta}} \sum_j 1_{\omega_j=\omega} \tilde{x}_j \Gamma_x^{e_j, j} [h_g | h_j, \eta, \tilde{x}_j] [v^{e_j, j+1}(h_g, p'_j) - v^{e_j, j+1}(h_b, p'_j)] \Pr[\eta | \omega_j, \tilde{x}_j] =$$

$$R \frac{1}{M_{\omega\eta}} \sum_j 1_{\omega_j=\omega} \tilde{x}_j \left( \sum_{h'} \Gamma_x^{e_j, j} [h' | h_j, \eta, \tilde{x}_j] \chi^{j+1}(h') [c^{e_j, j+1}(h', p'_j)]^{-\sigma_c} \right) \Pr[\eta | \omega_j, \tilde{x}_j]$$

where  $M_{\omega\eta} = \sum_j 1_{\omega_j=\omega} \Pr[\eta | \omega_j, \tilde{x}_j]$

- Note we use  $c^{e,i}(h, p)$  (a group average consumption) and  $v^{e,i}(h, p)$

## Moment conditions

### Average Health Transitions

- Moment conditions (M3) to (M4) identify the  $\lambda_{0\eta}^{ie}$
- $\forall \omega$  and  $X \in \{X_{L(\omega)}, X_{H(\omega)}\}$  we have

$$\begin{aligned} \tilde{\Gamma}(h_g|\omega, X) &= \sum_{\eta} \frac{1}{M_{\omega\eta X}} \sum_j 1_{\omega_j=\omega, \tilde{x}_j \in X} \left[ \lambda_{0\eta}^{ieh} + \lambda_{1\eta}^{ih} \frac{\tilde{x}_j^{1-\nu^h} - 1}{1 - \nu^h} \right] \Pr[\eta|\omega_j, \tilde{x}_j] \end{aligned}$$

where

- $M_{\omega\eta X} = \sum_j 1_{\omega_j=\omega, \tilde{x}_j \in X} \Pr[\eta|\omega_j, \tilde{x}_j]$
- $X_{L(\omega)} = \{x \leq \tilde{x}_{\text{med}}(\omega)\}$
- $X_{H(\omega)} = \{x > \tilde{x}_{\text{med}}(\omega)\}$

## Estimates of $\nu$ and $\lambda_1$

- Less curvature in health production than in consumption
  - ⇒ *ceteris paribus*, health expenditure shares increase with income  
(As in Hall, Jones (QJE 2007), but completely different identification)
  - But: in the cross-sectional data health expenditure shares unrelated to income
    - Poorer individuals have larger gains to leave bad health state
- Bad health outlook shock  $\eta_b$  increases return to money  
(especially so in good health state)

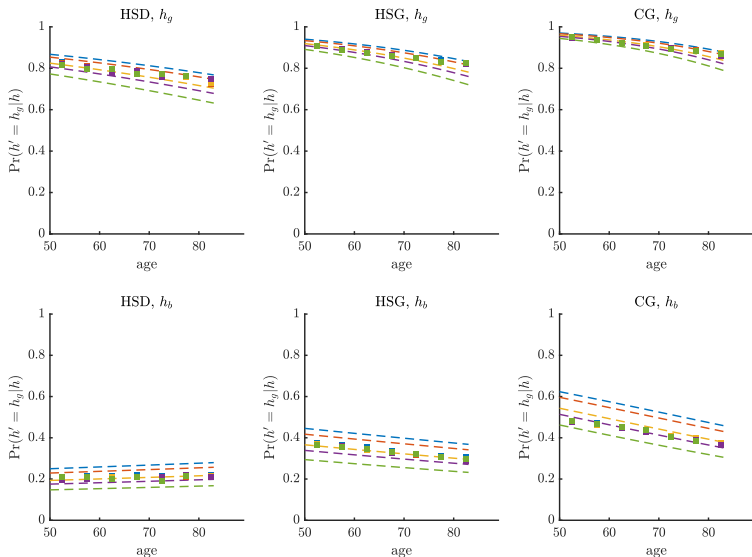
parameter	with $\pi = 0.5$
$\nu(h_g)$	1.2325 (0.022)
$\nu(h_b)$	0.8204 (0.034)
$\lambda_1(h_g, \eta_g)$	0.0466 (0.0087)
$\lambda_1(h_g, \eta_b)$	0.0912 (0.0169)
$\lambda_1(h_b, \eta_g)$	0.0019 (0.0006)
$\lambda_1(h_b, \eta_b)$	0.0022 (0.0007)

## Estimates of $\lambda_0$ : Take 1

- Our estimates generate health transitions that are consistent with
  - More educated have better transitions
  - Older have worse transitions
  - Useful medical spending predicts worse transitions in the panel
- ▷ BUT: *not enough separation of health transitions by wealth*
  - Given our estimates of  $\lambda_1$  and  $\nu$ , observed differences of OOP medical spending across wealth types are too small

# Health transitions: Wealth Matters in Data not in Model

*Data dashed and model dot each wealth quintile*



## Estimates of $\lambda_0$ : Take 2

- Let's allow the  $\lambda_0$  to depend on wealth
- We parameterize the age and wealth dependence of  $\lambda_{0\eta}^{iehp}$  as follows

$$\lambda_{0\eta}^{iehp} = \frac{\exp(L_{\eta}^{iehp})}{1 + \exp(L_{\eta}^{iehp})}$$

where  $L_{\eta}^{iehp} = a_{\eta}^{eh} + ap_{\eta}^{eh} \times (p - 3) + b_{\eta}^{eh} \times (i - 50)$

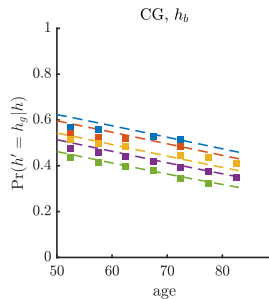
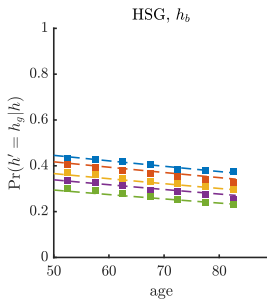
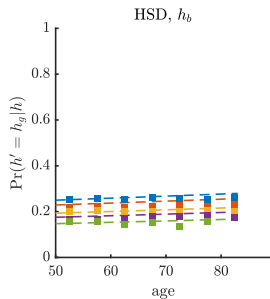
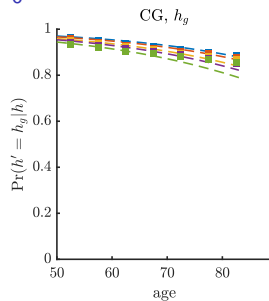
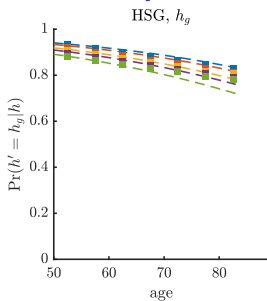
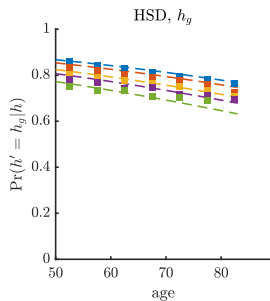
- We normalize  $\pi_{\eta} = 1/2$  and estimate

$$\theta_2 = \underbrace{\{a_{\eta}^{eh}, ap_{\eta}^{eh}, b_{\eta}^{eh}\}}_{\lambda_{0\eta}^{iehp}}, \lambda_{1\eta}^h, \nu^h, \sigma_{\epsilon}^2$$

(This is  $12+12+12+4+2+1 = 43$  parameters)

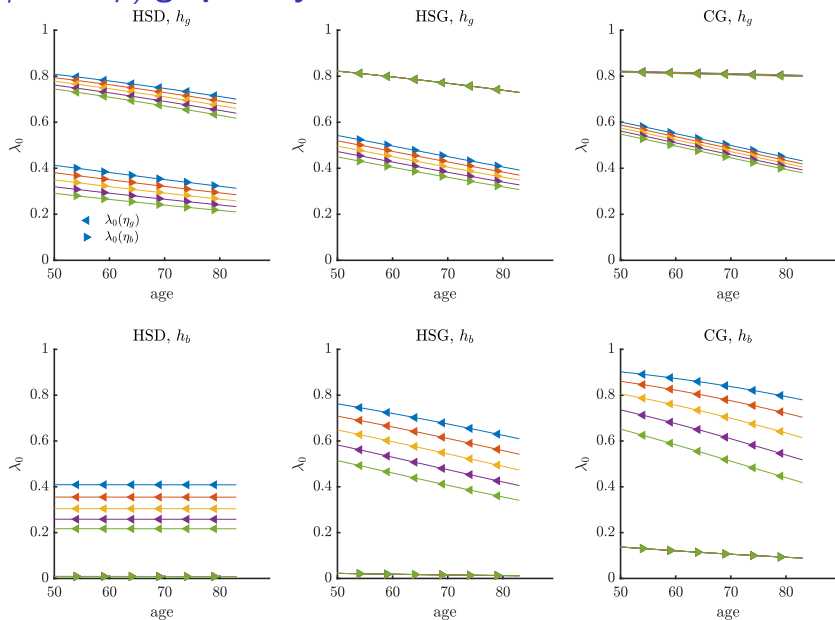
- Now: Wealthier experience better health transitions

# Health transition with wealth dependent $\lambda^p$





# $\lambda_0(\eta, i, e, h, \rho)$ graphically



## So what to do about wealth-dependent transitions?

### *Two strategies*

- ① Pose **unobserved types**: something that increases wealth AND health
  - Bad types dissave (cannot be done without fully solving the model).  
WHICH KILLS THE BEAUTY OF THE APPROACH!!!
  
- ② **Non-linear (concave) pricing**: difference in total health spending by wealth types is larger than in OOP
  - In preliminary estimates w/ MEPS data, the price of medical spending:
    - Declines with medical spending  $\Rightarrow$  concave pricing (copyments lower for more severe treatments)
    - Is lower for the less educated individuals (copyments lower in the public system)
    - Is higher in good health (copyments higher for preventive care)
  - But: MEPS lacks data on wealth

## Conclusions

# Conclusions

- We have identified preferences for health
  - Consumption is complement with health
  - Differential value of good health seems to be increasing with age.
  - Health is very valuable:
    - Back of the envelope calculation says that the better health of college educated than high school dropouts is worth 5 times the consumption of the latter group.
  
- Health technology
  - Health expenditures matter little
  - Wealth matters beyond health expenditures
    - Perhaps additional type differences
    - Perhaps concave pricing
    - Perhaps differential use of expenditures