# Non-Price Competition and Risk Selection Through Hospital Networks

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### Introduction

• Risk selection is a main concern in insurance markets.

• Health insurers may attempt to enroll healthy (profitable) instead of sick (unprofitable) patients (e.g., through prices, ads, networks).

• Risk selection reduces access to insurance and health care.

• Risk selection may also lead a market to unravel altogether.

# Research questions

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  - Basis for quantifying risk selection incentives is a demand model.
  - ► Supply model to measure how insurers' network choices respond.

- Better risk adjustment (gov. payments to insurers).
- Modify how insurers compete on premiums.

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- Study whether and how health insurers engage in risk selection through the design of their hospital networks.
  - Insurers risk-select by providing a narrow hospital network (i.e., fewer in-network hospitals) in services unprofitable patients need.

- What policies reduce the distortion in networks due to selection?
  - Better risk adjustment  $\rightarrow$  increases avg. network breadth by 28%.
  - Premium competition  $\rightarrow$  increases avg. network breadth by 30%.

# Contribution

- Document insurers' incentives to use narrow networks to risk-select.
  - Contributes to literature on risk selection mechanisms (Geruso et al., 2019; Aizawa and Kim, 2018).
- Selection on *multidimensional* service-level hospital networks.
  - ▶ Builds on Shepard (2022) who studies selection on *one* hospital.

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- Selection on *multidimensional* service-level hospital networks.
  - ▶ Builds on Shepard (2022) who studies selection on *one* hospital.
- Endogeneize hospital network breadth in a tractable way.
  - Related papers in this literature are Prager and Tilipman (2020); Ghili (2020); Ho and Lee (2019); Liebman (2018).

- Empirical setting is Colombia:
  - Contributory system (CR) (49% of population).
  - One national insurance plan provided by private insurers.
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  - Gov. risk adjustment formula is coarse. Expand

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2 Descriptive evidence

3 Model





# Data overview: sample of enrollees

- All covered by the CR in 2010-2011 (25 MM) and medical claims (650 MM).
- Continuous enrollment spells (9 MM) and claims (270 MM).
  - 1/3 are new enrollees in 2011. Expand

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- Keep 14 largest insurers. Account for 97% of enrollees.
- Market is a Colombian state (similar to MSA). 32 markets.



Figure: Insurers per market

## Data overview: services and networks

- Collapse 7,000 services codes into 58 categories ("services"). E.g:
  - Procedures in cardiac vessels.
  - Procedures in intestines.
  - Procedures in bones and joints.
  - Procedures in skull and brain.
  - Hospitalization.
  - Consultations.

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- Recover service-specific hospital networks from observed claims.
- Drop small providers. Final sample of "hospitals" represents 32% of total costs in the CR and 40% of total costs for the average insurer.



## Network breadth as a means of risk selection

- Narrow networks is one of the main reasons for dissatisfaction with an insurer in Colombia (based on surveys by the Ministry of Health).
- Object of interest is insurer *j*'s service-level hospital network breadth in market *m* and service *k*, **H**<sub>ikm</sub>.

# Network breadth as a means of risk selection

- Narrow networks is one of the main reasons for dissatisfaction with an insurer in Colombia (based on surveys by the Ministry of Health).
- Object of interest is insurer *j*'s service-level hospital network breadth in market *m* and service *k*, **H**<sub>jkm</sub>.
- *H<sub>jkm</sub>* ∈ [0, 1] is the fraction of hospitals in market *m* that provide service *k* that are covered by insurer *j*.
- Simplicity of *H<sub>jkm</sub>* allows for tractability. Limitation is treating hospital quality as constant.

Distribution of Hikm

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#### Conclusions

# Descriptive facts

• Risk selection incentives exist at the service level:

Average cost of patients that require complex services are almost 3 times higher than the average reimbursement. Expand

• Network breadth tends to be much smaller for services that are unprofitable. Expand

- Consumers respond to network breadth choices:
  - Women are more likely to have a baby the broader is the network for delivery services. Expand

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$$u_{ijm} = \beta_i \sum_{k} \underbrace{q_{\theta km}}_{\substack{\text{claim} \\ \text{prob.}}} \underbrace{H_{jkm}}_{\substack{\text{network} \\ \text{breadth}}} -\alpha_i \underbrace{c_{\theta jm}(H_{jm})}_{\substack{\text{OOP} \\ \text{costs}}} + \phi_j + \varepsilon_{ijm}$$

$$\beta_i = (x_i \ y_i)'\beta$$
$$\alpha_i = x'_i \alpha$$

- θ = sex, age group, diagnosis (cancer, cardio, diabetes, renal, other, 2 or more diseases, no diseases).
- $x_i = \text{sex}$ , age, diagnosis, location.  $y_i = \text{income group}$ .
- k is a service.
- $q_{\theta km}$  is the prediction of a logistic regression, off-line. Variation in q

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- Consumers face a cost-coverage trade-off.
- The magnitude of this trade-off will vary with health status.
- Allows for healthy individuals to be screened by narrow networks.
- ▶ OOP costs are endogenous to *H<sub>jkm</sub>*.

# Demand identification

- Preference for network breadth uses exogenous variation in market demographics.
  - Little concern over insurers targeting demographics with their networks.

• Marginal disutility for OOP costs uses exogenous variation in service reference prices in a control function approach. Expand



# Willingness-to-pay varies with health status

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- Significant disutility from OOP costs. Avg. elasticity = -0.51.
- Strong preferences for broader networks in services they need.
- Insurers can avoid these patients by providing narrow networks.

| Diagnosis         | Willingness-to-pay |
|-------------------|--------------------|
| Cancer            | 22.0               |
| Cardiovascular    | 7.0                |
| Diabetes          | 4.8                |
| Renal             | 5.3                |
| Other             | 14.8               |
| $\geq$ 2 diseases | 6.7                |
| Healthy           | 2.1                |
|                   |                    |

Demand model fi

$$\log(AC_{\theta jm}(H_{jm})) = \tau_0 \underbrace{\left(\sum_{k} q_{\theta km} A_k\right)}_{k} + \tau_1 \underbrace{\left(\sum_{k} q_{\theta km} H_{jkm}\right)}_{\text{network breadth}} + \frac{1}{2K_m} \tau_2 \underbrace{\sum_{k} \sum_{l \neq k} q_{\theta km} q_{\theta lm} H_{jkm} H_{jlm}}_{\text{scope}} + \lambda_{\theta} + \eta_m + \delta_j$$

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- $A_k$  is the government's reference price.

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- Insurers observe  $q_{\theta km}$ .
### Insurer average costs per enrollee

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- Approximation to a price bargaining equilibrium.
- $A_k$  is the government's reference price.
- Insurers observe  $q_{\theta km}$ .
- Scope economies: insurers that offer broad networks in one service, tend to offer broad networks in other services.

### Average cost regression

• Identification relies on variation in market demographics across markets, within insurer.



Figure: Predicted average cost

### Insurer total average costs

Take one market:

 $\sum_{\theta} \underbrace{AC_{\theta j}(H_j)}_{\xi \theta j} \underbrace{s_{\theta j}(H)N_{\theta}}_{\xi \theta j}$ 

- Network breadth affects average costs directly through  $AC_{\theta_i}$ .
- (Selection:) Network breadth affects the composition of consumer types in demand through s<sub>θi</sub>N<sub>θ</sub>, where s<sub>θi</sub> is the choice probability.

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- (Selection:) Network breadth affects the composition of consumer types in demand through s<sub>θj</sub>N<sub>θ</sub>, where s<sub>θj</sub> is the choice probability.
- Decomposition tells us how important is adverse selection vs. cost heterogeneity in generating observed data patterns.

Insurer competition

Let  $\pi_{ijm}(H_m, \theta)$  be insurer j's annual per-enrollee profit. Depends on *own* and *rival* network breadth,  $H_m = \{H_{jm}\}_{j=1}^{\#\mathcal{J}_m}$ , where  $H_{jm} = \{H_{jkm}\}_{k=1}^{\#\mathcal{K}_m}$ 

$$\pi_{ijm}(H_m, \theta) = (\underbrace{R_{\theta m}}_{\substack{\text{govmt} \\ \text{transfer} \\ + \text{copays}}} - \underbrace{(1 - r_i)}_{\substack{\text{average} \\ \text{cost}}} \underbrace{AC_{\theta jm}(H_{jm})}_{\substack{\text{average} \\ \text{cost}}} \underbrace{S_{ijm}(H_m)}_{\substack{\text{choice} \\ \text{prob.}}}$$

• Nash equilibrium. Insurers choose networks to maximize:

 $\Pi_{jm}(H_m) = \text{short-run profit} + \text{long-run profit} - \text{network formation cost}$ 

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• Steady state Nash equilibrium. Insurers choose networks to maximize:

$$\mathbf{I}_{jm}(H_m) = \sum_{\theta,m} \left( \underbrace{\pi_{ijm}(H_m, \theta) N_{\theta m}}_{\text{short-run profit}} + \sum_{s=t+1}^{T} \zeta^s \sum_{\theta'} (1 - \rho_{\theta'}) \mathcal{P}(\theta'|\theta) \pi_{ijm}(H_m, \theta') N_{\theta'm} \right)$$

$$\lim_{\text{long-run profit}} - \underbrace{\sum_{k} (\omega H_{jkm} + \xi_{jkm}) H_{jkm}}_{\text{network formation cost}}$$

• Individuals experience infinite inertia (recall switching rate is 0.06%).

• 
$$\xi_{jkm} = \xi_j + \xi_k + \xi_m + \vartheta_{jkm}$$

FOC at an interior solution  $H_{jkm} \in (0, 1)$ :

$$MVP_{jkm}(H_m) = 2\omega H_{jkm} + \xi_j + \xi_k + \xi_m + \vartheta_{jkm}$$

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- *MVP<sub>jkm</sub>* is the marginal variable profit.
- Insurers *internalize* the cost-coverage trade-off from demand.

$$c_{\theta jm} = \mu_y A C_{\theta jm} (H_{jm}) + \epsilon_{\theta jm}$$

•  $H_{jkm}$  is observed together with  $\vartheta_{jkm}$ .

Relation between r and  $\mu$ 

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• *MVP<sub>jkm</sub>* is the marginal variable profit.



• Firms might resolve these trade-offs differently *even if they have identical network formation costs.* 

# Network formation cost identification and estimation

#### Identification:

- Variation in average claim probabilities across markets.
- Network breadth in 2010.
- Estimate FOC in 4 largest markets with no corners.

First-stage Out-of-

Out-of-sample fit

# Table: Model of insurer network formation costs

| $log(MVP_{jkm})$   | Coefficient | Std. Error |  |
|--------------------|-------------|------------|--|
| Network            | 3.41***     | 0.07       |  |
| Insurer FEs        |             |            |  |
| EPS001             | -0.79***    | 0.04       |  |
| EPS002             | -0.14***    | 0.04       |  |
| EPS003             | -0.50***    | 0.04       |  |
| EPS005             | -1.37***    | 0.04       |  |
| EPS010             | 0.38***     | 0.04       |  |
| EPS013             | -0.37***    | 0.04       |  |
| EPS016             | -0.34***    | 0.04       |  |
| EPS017             | -0.80***    | 0.04       |  |
| EPS018             | -0.53***    | 0.04       |  |
| EPS037             | (ref)       | (ref)      |  |
| First stage F-stat | 77-         | 4.5        |  |
| N                  | 2,262       |            |  |
| $R^2$              | 0.          | 97         |  |

Note: Includes insurer, market, and service fixed effects.

# Model-based evidence of adverse selection

• Insurer average and marginal costs are positively correlated with consumer willingness-to-pay for network breadth.

- *Decomposition exercise*: suppose an insurer deviates and increases network breadth by 10%.
  - ► Adverse selection explains <u>48%</u> of variation in total costs.
  - Cost heterogeneity explains the rest.

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#### Conclusions

# The effect of risk adjustment of network breadth

- Simulate two counterfactual scenarios:
  - Eliminate risk adjustment.
  - Improve risk adjustment.

- For simplicity, estimate counterfactuals with data only from the capital, Bogotá.
  - ▶ Around 1/3 of all enrollees to the contributory regime live in Bogotá.
  - Has presence of all insurers.

# No risk adjustment

• Per-capita transfer equals national base transfer times adjustment factor to match observed short-run gov spending:

$$R_{\theta m}^{cf} = \lambda \times R, \quad \forall (\theta, m)$$

• Prediction: Eliminating risk adjustment should exacerbate risk selection incentives and reduce network breadth.

### Improved risk adjustment

• Counterfactual risk-adjusted transfer is:

$$R_{\theta m}^{cf} = \lambda \times a_m \times 360 \times \frac{\sum_{\theta(i)=\theta} T_i}{\sum_{\theta(i)=\theta} b_i}$$

*T<sub>i</sub>* is total cost, *b<sub>i</sub>* is number of days enrolled in the year, *a<sub>m</sub>* is a market multiplier, λ is adjustment factor to match gov spending.

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- *T<sub>i</sub>* is total cost, *b<sub>i</sub>* is number of days enrolled in the year, *a<sub>m</sub>* is a market multiplier, λ is adjustment factor to match gov spending.
- *Eliminates demand-side incentives to risk-select.* Shows relative importance of adverse selection vs. cost heterogeneity.
- Prediction: Improving risk adjustment should reduce risk selection incentives and increase network breadth.

Observed risk adjustment is better than no risk adjustment

Table: Changes in networks, costs, and welfare under alternative risk adjustment

| Variable                          | No RA | Improved RA |             |           |
|-----------------------------------|-------|-------------|-------------|-----------|
|                                   |       | 7 diseases  | 30 diseases | "Perfect" |
| Panel A. Overall                  |       |             |             |           |
| Avg. network breadth              | -6.7  | 4.6         | 10.9        | 28.0      |
| Avg. cost per enrollee            | -0.9  | 1.1         | 3.7         | 3.0       |
| Consumer welfare (healthy)        | -2.1  | 2.8         | 9.9         | 7.7       |
| Consumer welfare (sick)           | -3.3  | 3.4         | 10.7        | 11.1      |
| Panel B. Avg. network per service |       |             |             |           |
| Abdominal wall                    | -25.7 | 19.0        | 37.9        | 106.7     |
| Imaging, lab, consultation        | -6.2  | 2.9         | 10.0        | 21.3      |
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# The effect of premiums on network breadth

- Insurers compete simultaneously in premiums and networks.
- Insurers can discriminate premiums along sex, age, and income.

# The effect of premiums on network breadth

- Insurers compete simultaneously in premiums and networks.
- Insurers can discriminate premiums along sex, age, and income.
- In the observed scenario:

$$c_{ heta jm} = \text{Coins}_{ heta jmk} + \text{Copay}_{ heta m} + \underbrace{\text{Tax}}_{1/3 \text{ of total taxes}}$$

• In counterfactual, let  $\theta = (\theta_1, \theta_2)$ , where  $\theta_2$  are diagnoses.

$$\mathsf{Coins}_{ heta jm} + \mathsf{Copay}_{ heta m} + \mathsf{Tax} + (1/3) imes ilde{P}_{ heta_1 jm}$$

• Calibrate  $\alpha$  to (roughly) match average elasticity from other papers: Abaluck and Gruber (2011) -1.17; Shepard (2022) -1.48.

Profit function

| Variable     |                  | Low price sens. | Med price sens. |
|--------------|------------------|-----------------|-----------------|
| Sex          | Female           | 89              | 84              |
|              | Male             | 165             | 158             |
| Age group    | <1               | _               | _               |
|              | 1-4              | _               | _               |
|              | 5-14             | 145             | 139             |
|              | 15-18            | 151             | 147             |
|              | 19-44            | 107             | 104             |
|              | 45-49            | 86              | 80              |
|              | 50-54            | 73              | 67              |
|              | 55-59            | 184             | 175             |
|              | 60-64            | 139             | 132             |
|              | 65-69            | 124             | 117             |
|              | 70-74            | 135             | 128             |
|              | $\geq$ 75        | 125             | 119             |
| Income group | $< 2 \times MMW$ | 203             | 196             |
|              | [2,5] × MMW      | 51              | 46              |
|              | $> 5 \times MMW$ | —               | —               |

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|--------------|---|-----------------|-----------------|
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| Income group | ≥75   | 125             | 119             |
|              | < 2 × MMW   | 203             | 196             |
|              | [2,5] × MMW   | 51              | 46              |
|              | > 5 × MMW   | —               |                 |

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## Premiums and networks are substitutes for risk selection

| Variable                          | Low price sens. | Med price sens. |
|-----------------------------------|-----------------|-----------------|
| Panel A. Overall                  |                 |                 |
| Avg. network breadth              | 31.6            | 27.7            |
| Total avg. cost                   | 4.2             | 1.5             |
| Total revenue                     | 21.5            | 18.4            |
| Consumer welfare (healthy)        | 5.3             | -3.5            |
| Consumer welfare (sick)           | 4.2             | -9.2            |
| Avg. premium elasticity           | -0.9            | -1.2            |
| Panel B. Avg. network per service |                 |                 |
| Abdominal wall                    | 71.1            | 67.1            |
| Imaging, lab, consultation        | 20.4            | 11.1            |
| Hospital admission                | 21.9            | 14.6            |

Table: Changes in networks, costs, and welfare under premium deregulation

*Note*: Table presents percentage change in counterfactual relative to observed scenario. Baseline average network breadth equals 0.38.

## Conclusions

- Study how insurers use their hospital networks to risk-select.
- In a setting where hospital networks are service specific:
  - Consumers choose insurers with broad networks in services they need.
  - Insurers choose their hospital networks per service to select the most profitable consumers.
- Better risk adjustment increases network breadth by 28%, holding government spending fixed.
- Premiums and hospital networks are substitutes for risk selection.
- Zero premiums lead to narrow networks.

# Thank you

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## Risk adjustment

- Risk-adjusted capitated payments from gov. to insurers.
- Payments equal average health care cost per risk pool.
- Ex-ante risk adjustment:
  - Paid at the beginning of every year.
  - Risk pool is a combination of sex, age group, location.
- Ex-post risk adjustment:
  - Paid at the end of every year.
  - Insurers with above-average share of patients with certain diseases receive payments from those with below-average share.



#### New enrollees

- People who move from subsidized (SR) to contributory system.
- People who age into the contributory system.
- Insurers in the CR participate in the SR.
- 1/6 of my sample moved from only one insurer in the SR to the CR.
- People with 3 continuous months of non-payment of taxes would be disenrolled and information removed from system → "fresh start".



## Service coverage within hospital



Figure: Fraction of services covered per hospital (zero is 60%)

### $H_{jkm}$ varies either because of selection or cost differences



#### Figure: Distribution of network breadth

Note: Distribution of service-level network breadth conditional on four largest markets.

• Variation: 30% insurer, 10% service, 4% market. Back

## Do risk selection incentives exist? (Geruso et al., 2019)



Note: Dots are services weighted by number of enrollees who make claims for the service. One enrollee can appear in multiple dots. Enrollees who make no claims are not represented in this figure.

## Distribution of health care costs



Figure: Health care cost by risk-adjusted transfer

## $H_{jkm}$ covaries with service profitability – selection story



#### Figure: Correlation between network breadth and service profitability

Note: Dots are services weighted by number of enrollees who make claims for the service. One enrollee can appear in multiple dots. Enrollees who make no claims are not represented in this figure.

## Do consumers respond to $H_{jkm}$ ?

- Claim probability is positively correlated with network breadth.
  - Women in childbearing ages choose insurers with broad networks for delivery.
  - ► Higher likelihood of dialysis and chemo claims the broader the networks for renal disease and cancer treatment.
- People switch towards insurers with broad networks after health shock (.06% of current).
  - The newly diagnosed with arthritis switch to insurer with broad network for procedures in bones and joints.
- Insurers with broad networks imply higher out-of-pocket costs.



## Selection into moral hazard

#### Table: Service-specific network breadth and types of claims

|  | (1) Current | (2) Full  |
|--|-------------|-----------|
| (1) Any childbirth claim               |             |           |
| H <sub>im</sub> Delivery               | 0.02***     | 0.01***   |
|  | (0.001)     | (0.001)   |
| Ν                                      | 1,085,206   | 3,078,555 |
| (2) Any dialysis claim                 |             |           |
| <i>H<sub>jm</sub></i> Dialysis         | 0.03***     | 0.03***   |
|  | (0.004)     | (0.003)   |
| Ν                                      | 83,768      | 120,330   |
| (3) Any antirheumatic drug claim       |             |           |
| <i>H<sub>jm</sub></i> Bones and Joints | 0.002       | 0.002**   |
|  | (0.001)     | (0.001)   |
| Ν                                      | 102,612     | 156,385   |
| (4) Any chemotherapy claim             |             |           |
| H <sub>jm</sub> Therapy                | 0.003*      | -0.002    |
|  | (0.002)     | (0.001)   |
| N                                      | 439,176     | 785,727   |

Note: Each regression is conditional on the sample of individuals who received a diagnosis during 2010. All regressions include market fixed effects and control for sex and age group. Robust standard errors in parenthesis. \*\*\*p<0.01, \*\*p<0.05, \*p<0.1.

## Where do people switch after receiving a diagnosis?

Table: Insurer choice among switchers with changes in health status

|  | Insurer choice |
|--|----------------|
| (1) Women in childbearing ages                       |                |
| $H_{im}^{2010} - H_{i'm}^{2011}$ Delivery            | -2.77***       |
|  | (0.12)         |
| Ν  | 14,958         |
| (2) Additional diagnosis of renal disease            |                |
| $H_{im}^{2010} - H_{i'm}^{2011}$ Dialysis            | -1.51*         |
| , , , , , , , , , , , , , , , , , , ,                | (0.84)         |
| Ν  | 40             |
| (3) Additional diagnosis of cancer                   |                |
| $H_{im}^{2010} - H_{i'm}^{2011}$ Therapy             | -3.23***       |
| , , , , , , , , , , , , , , , , , , ,                | (0.37)         |
| Ν  | 1,658          |
| (5) Newly diagnosed                                  |                |
| $H_{im}^{2010} - H_{i'm}^{2011}$ Hospital admissions | -1.94***       |
| . ,  | (0.21)         |
| Ν  | 5,787          |

Note: Conditional logit estimated on the sample of switchers with a new diagnosis. The main explanatory variable is the difference in network breadth for delivery services between the incument insurer j and all other insurers j'. Robust standard errors in parenthesis. \*\*\* p < 0.01, \*\* p < 0.05, \*p < 0.1. Back

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## Variation in OOP costs



Figure: Distribution of OOP costs as percentage of monthly minimum wage

• Explained variation: 66% consumer types, 33% insurer-market.

## OOP costs and network breadth



Figure: Correlation between OOP costs and network breadth

## Relation between $r_y$ and $\mu_y$

- If OOP costs were only coinsurance payments, then  $r_y = \mu_y$ .
- OOP costs include other components that the insurer does not cover (taxes and copays), so we can expect r<sub>y</sub> ≥ μ<sub>y</sub>

|                  | (                | Out-of-pocket cost |                  |  |
|------------------|------------------|--------------------|------------------|--|
|                  | (1)<br>< 2 × MMW | (2)<br>[2,5] × MMW | (3)<br>> 5 × MMW |  |
| $AC_{\theta jm}$ | 0.083***         | 0.143***           | 0.209***         |  |
|                  | (0.0004)         | (0.0002)           | (0.003)          |  |
| Constant         | 0.036***         | 0.061***           | 0.182***         |  |
|                  | (0.0002)         | (0.0001)           | (0.002)          |  |
| N                | 162,464          | 334,961            | 2,575            |  |
| $R^2$            | 0.19             | 0.63               | 0.68             |  |

Table: Pass-through of average costs to out-of-pocket costs

### Game

• Players: insurers,  $j \in \mathcal{J}_m$ 

• Payoffs: 
$$\Pi_{jm}(H_m, \vartheta_{jkm}) \in \mathbb{R}$$
,  
where  $H_m = \{H_{jm}\}_{j=1}^{\#\mathcal{J}_m}$  and  $H_{jm} = \{H_{jkm}\}_{k=1}^{\#\mathcal{K}_m}$ 

- $\bullet~Strategies:~\mathbb{R} \rightarrow [0,1]^{58}$
- Information:  $\vartheta_{jkm}$  is private information. Rest is common knowledge  $(q_{jkm}, \rho_{\theta}, \mathcal{P}(\theta'|\theta)).$

### Reference service prices

• In 2005 the Colombian government published a list of reference prices for each service in the national plan.

• Hospitals are reimbursed with these prices in three situations: terrorist attacks, car accidents, natural disasters.

• Reference prices were not meant to guide insurer-hospital negotiations. But insurers use them as a starting point.

#### Reference service prices



Figure: Correlation between average negotiated price and reference price

## Control function approach

• Problem: OOP costs may be correlated with unobserved quality.

• *Solution*: use the government's reference price per service as instrument.

• *Exclusion*: reference prices shift supply, unobserved by consumers.

• *Relevance*: reference prices affect OOP costs through their effect on insurers' average cost per enrollee.

## Control function approach

• First stage:

$$c_{\theta jm} = \beta_0 + \beta_1 (1 - r_i) \sum_m q_{\theta km} A_k + \lambda_{\theta} + \delta_j + \eta_m + \varphi_{\theta jm}$$

• Obtain  $\hat{\varphi}_{\theta jm}$ , standardize at the market level, then estimate the following:

$$u_{ijm} = \beta_i^D \sum_{k} q_{\theta km} H_{jkm} - \alpha_i c_{\theta jm} (H_{jm}) + x_i' \hat{\varphi}_{\theta jm}^z + \phi_j + \varepsilon_{ijm}$$

• Conditional on  $\hat{\varphi}^{z}_{\theta jm}$ ,  $c_{\theta jm}$  is orthogonal to  $\varepsilon_{ijm}$ .

## Identification threats in demand

Network breadth may be correlated with unobserved insurer quality. Robustness checks:

- Include other insurer quality measures (enrollee satisfaction score, avg. wait times).
- Include a star hospital coverage indicator.
- Subsample of markets without star hospitals.

- Network breadth assumes hospital quality as a constant.
- Problematic if it matters which hospitals are included and not just how many (e.g, star hospitals).

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  - Model specification.
    - Arises if network breadth is negatively correlated with hospital quality this is not the case.
  - Ø Bias.
    - Arises if there is significant variation in hospital quality within service robustness checks suggest otherwise.

## Identification threats in cost

• There may be unobserved cost variation within consumer type.

- Adverse selection can bias coefficients in average cost function.
  - Robustness check using patient-level data.

# Variation in $q_{\theta km}$



Figure: Variation in service claim probability

### Demand model fit

| Insurer | Observed | Predicted |
|---------|----------|-----------|
| EPS001  | 2.06     | 2.05      |
| EPS002  | 7.46     | 7.52      |
| EPS003  | 4.19     | 4.14      |
| EPS005  | 4.33     | 4.36      |
| EPS008  | 3.88     | 3.88      |
| EPS009  | 2.11     | 2.09      |
| EPS010  | 6.38     | 6.38      |
| EPS012  | 1.19     | 1.18      |
| EPS013  | 16.10    | 16.17     |
| EPS016  | 19.78    | 19.78     |
| EPS017  | 6.41     | 6.47      |
| EPS018  | 4.16     | 4.20      |
| EPS023  | 2.38     | 2.38      |
| EPS037  | 19.59    | 19.39     |
|         |          |           |

#### Table: National market shares

## Marginal variable profits

Table: Summary statistics of marginal variable profits per insurer

| Insurer | MVP           |
|---------|---------------|
| EPS001  | 272 (1,143)   |
| EPS002  | 829 (3,095)   |
| EPS003  | 400 (1,551)   |
| EPS005  | 222 (864)     |
| EPS010  | 894 (3,067)   |
| EPS013  | 717 (2,331)   |
| EPS016  | 1,276 (4,166) |
| EPS017  | 619 (3,339)   |
| EPS018  | 571 (2,224)   |
| EPS037  | 1,103 (3,533) |
|         |               |

#### First-stage

| H <sub>jkm</sub>                                     | Coefficient | Std. Error |
|--|-------------|------------|
| $H_{ikm}^{t-1}$                                      | 0.76***     | 0.01       |
| $\overline{q}_{female,k,m}$                          | 33.93***    | 8.45       |
| $\overline{q}_{healthy,k,m}$                         | 14.30***    | 4.21       |
| $\overline{q}_{age 19-44,k,m}$                       | -55.33***   | 13.63      |
| $H_{jkm}^{t-1} 	imes \overline{q}_{age \ 19-44,k,m}$ | 0.16***     | 0.05       |
| N  | 2,262       |            |
| F-stat   | 774.45      |            |

Table: First stage regression of network breadth

Note: Includes insurer, market, and service fixed effects. Robust standard errors and first-stage F-statistic reported. \*\*\* p<0.01, \*\*p<0.05, \*p<0.0.

## Network formation cost results

Table: Predicted average total network formation cost per market

| Insurer | (1) Total | (2) % |
|---------|-----------|-------|
| EPS001  | 2,428     | 57    |
| EPS002  | 9,738     | 92    |
| EPS003  | 4,430     | 85    |
| EPS005  | 2,557     | 82    |
| EPS010  | 7,807     | 61    |
| EPS013  | 8,457     | 84    |
| EPS016  | 15,139    | 82    |
| EPS017  | 7,330     | 69    |
| EPS018  | 6,288     | 61    |
| EPS037  | 13,399    | 74    |

## Out-of-sample fit



Figure: Comparison of model's predictions to public income statements

## Insurer profit function with premiums

• Let 
$$P_m = \{\{P_{\theta_1 jm}\}_{\theta_1}\}_{j=1}^{\#\mathcal{J}_m}$$
. The annual per-enrollee profit is:  

$$\pi_{ijm}(H_m, P_m, \theta) = (\underbrace{R_{\theta m}}_{\substack{\text{govmt} \\ \text{transfer} \\ + \text{ copays}}} + \underbrace{P_{\theta_1 jm}}_{\text{premium}} - \underbrace{(1 - r_i)}_{\substack{\text{average} \\ \text{cost}}} \underbrace{AC_{\theta jm}(H_{jm})}_{\substack{\text{average} \\ \text{cost}}} \underbrace{s_{ijm}(H_m, P_m)}_{\substack{\text{choice} \\ \text{prob.}}}$$

- FOC w.r.t to  $P_{\theta jm}$  defines a fixed point in premiums.
- FOC w.r.t to H<sub>jkm</sub> defines a fixed point in networks.
- Simulation is a nested fixed point.

