

# Moral Hazard Induced Unraveling

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## Motivation:

- Many of the recent insurance expansions in US have relied on “managed competition.”
  - ▶ Private insurers.
  - ▶ Limit price discrimination.
    - Community rating.
  - ▶ Increase affordability.
    - Means-tested subsidies for premiums and out-of-pocket (OOP) payments.
  - ▶ Medicare Part D, Medicare Advantage, and the Affordable Care Act Health Insurance Exchanges all have these features.

## Motivation:

- Governments offer OOP subsidies in a tradeoff of affordability and the inefficiencies from moral hazard.
- It is very common for governments to intervene in markets with adverse selection.
  - ▶ However, it is usually thought the government has less of a role to play in “correcting” moral hazard.
  - ▶ This is because there typically isn’t an externality for moral hazard.
  - ▶ Community rating changes that though!

## Research Questions:

- Can means-tested, OOP subsidies unravel insurance markets with community rating?
- Is this happening on the ACA Exchanges?
- How bad is the welfare loss?

# Research Questions:

- **Can means-tested OOP subsidies unravel insurance markets with community rating?**
  - ▶ Yes. The combination of subsidies and community rating mimics adverse selection.
  - ▶ Shown graphically.
- **Is this happening on the ACA Exchanges?**
  - ▶ Yes. Reimbursing the MH component of OOP subsidies would lower premiums by around \$1000 and increase unsubsidized enrollment.
  - ▶ Medicaid Expansion Difference-in-Differences.

# Research Questions:

- **How bad is the welfare loss?**
  - ▶ About \$50 per potential market participant.
  - ▶ 25% of the cost of adverse selection.
  - ▶ Structural Model.

## Brief Background:

- Our empirics focus on the ACA, so I'll use that language.
  - ▶ Shown graphically, have a general theory in the paper.
- ACA subsidies on two margins: OOP and Premiums.
  - ▶ Premium Subsidies:
    - Premiums are capped at 2% of income for everyone under 400% of the FPL.
    - In practice, this is binding for most people.
  - ▶ OOP Subsidies:
    - People under 250% of the FPL have their cost-sharing subsidized.
    - People 100% - 150% FPL pay just \$0.06 on the dollar for healthcare.

# Example: Means-Tested OOP Subsidies

	<b>Standard Silver – No CSR</b>	<b>CSR Plan for 201-250% FPL</b>	<b>CSR Plan for 151-200% FPL</b>	<b>CSR Plan for up to 150% FPL</b>
Actuarial Value	70% AV	73% AV	87% AV	94% AV
Deductible (Individual)	\$7,150	\$4,500	\$800	\$250
Maximum OOP Limit (Individual)	\$7,350	\$5,700	\$1,700	\$550
Inpatient hospital	30% (after deductible)	30% (after deductible)	10% (after deductible)	10% (after deductible)
Physician visit	\$70	\$30	\$10	\$5

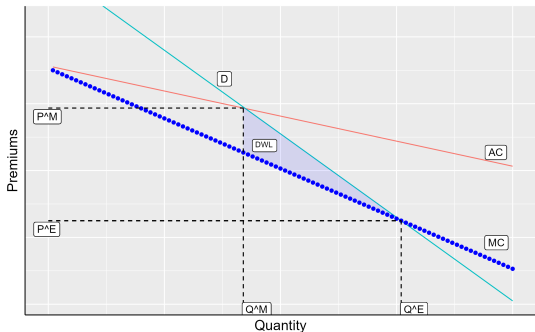


# Theory:

- (1) Moral Hazard means OOP subsidies induce higher costs.
- (2) Community rating spreads this increased cost to unsubsidized enrollees.
- (3) Premium subsidies ensure that *only* the unsubsidized feel the cost increase.
  - We can show how this works in the graphical framework of Einav and Finkelstein (2011).

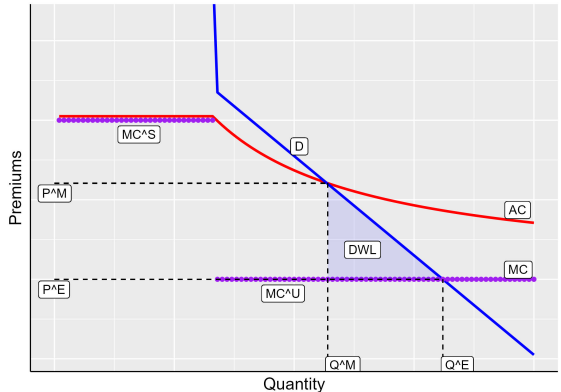
# Theory: Unraveling

- Brief reminder of Einav and Finkelstein Model.
- Adverse selection means  $MC$  curve is downward sloping.
  - ▶ Higher cost = higher demand
- Downward sloping  $MC$  curve means  $AC > MC$ .
- Inefficiency comes from under-insurance.



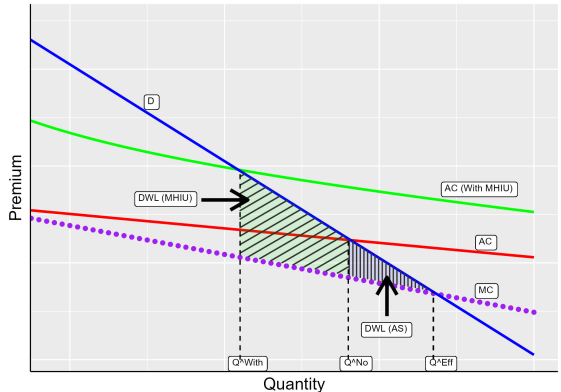
# Theory: Unraveling

- Same base MC (no selection).
- OOP subsidies increase MC for some consumers.
- Premium subsidies move them to the “left”.
- Makes AC curve slope down, mimicking adverse selection.



# Theory: Unraveling

- What if we add adverse selection?
- The problem gets worse!
- This is because the people who are kicked out have a higher risk premium.



# Welfare Estimation

Assume linear demand and cost functions (Einav, Finkelstein, and Cullen (2010)):

$$D(P) = \alpha + \beta * P$$

$$AC(P) = \gamma + \delta * P + \sigma * \mu$$

- $\delta$  and  $\gamma$  are the adverse selection parameters.
- $\mu$  is the additional moral hazard cost.
- $\sigma$  is the share of enrollees that get subsidies.
- Equilibrium  $\implies D(P) = AC(P)$ .
  - ▶ We also allow for markups later.

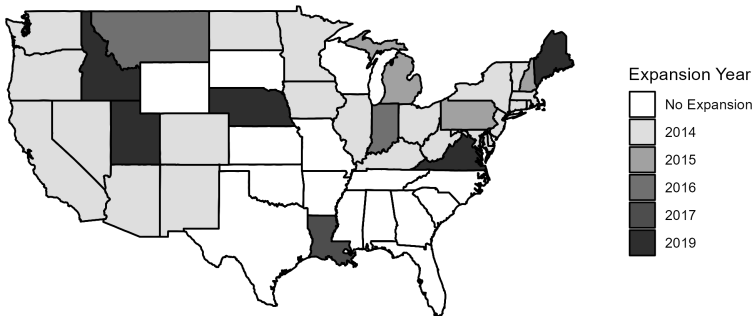
# Welfare Estimation

- $\delta$  and  $\mu$  we take from the literature.
  - ▶  $\delta = .155$  (Einav, Finkelstein, and Cullen (2010))
  - ▶  $\mu = \$721$ 
    - Determined by the elasticity from the RAND Health Insurance Experiment.
  - ▶ Use these to back out  $\gamma$ .
- $\sigma$  we observe in the data.
- $\alpha$  and  $\beta$  we need to estimate.

# Estimating $\alpha$ and $\beta$

- We need something that changes the share of subsidized enrollees, which changes premiums, but doesn't otherwise impact demand by the unsubsidized.
- The ACA's Medicaid Expansion fits this perfectly.
  - ▶ If you are eligible for Medicaid, you don't get exchange subsidies.
  - ▶ So Medicaid Expansion mechanically reduces the percentage of subsidized enrollees by about 40%.

# Affordable Care Act: Medicaid Expansion





## Model: Difference-in-Differences

- Borusyak et al. (2021) imputed diff-in-diff approach.

For (1) share HIX enroll 150% FPL, (2) premiums, (3) HIX enroll 400+% FPL, and (4) uninsured 400+% FPL, we estimate,

$$y_{st}^0 = x_{st}'\alpha + \theta_s + \tau_t + \varepsilon_{st} \quad (1)$$

$$y_{st}^1 - \widehat{y}_{st}^0 = \beta_1 \text{Expand}_{st} + \gamma_{st} \quad (2)$$

- (1) uses non-treated units and (2) uses all units.
- $x_{st}$  → controls
- $\theta_s$  and  $\tau_t$  are state (or rating-area) and year fixed effects.
- Combine estimates on HIX premiums and HIX enrollment as an IV to causally estimate demand.

# Data:

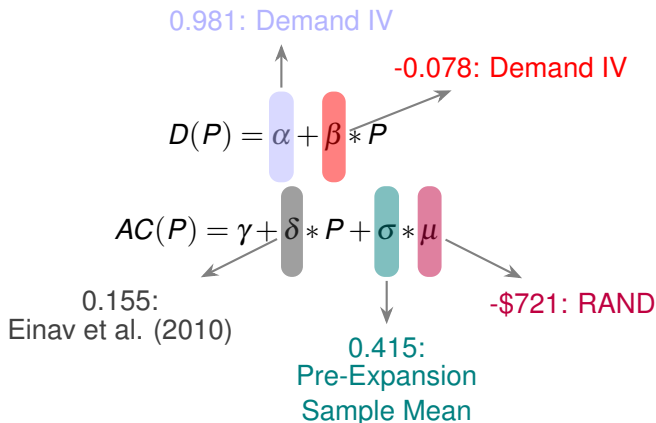
- HIX Compare: 2014-2017
  - ▶ Premiums for 27 years olds at the rating area, plan, carrier, and metal level.
  - ▶ Sample: 275 rating areas with 59,013 plans.
  - ▶ Utilize 2015-2017 expanders.
  - ▶ Avg. annual premium: \$3,400.
- HIX Open Enrollment Period (OEP) Data: 2015-2017
  - ▶ Number enrollees in a given rating area based on income.
  - ▶ 40% enrolled on HIX → 100-150% FPL.
- American Community Survey (ACS): 2012-2017
  - ▶ Health insurance takeup data:
    - Uninsured Rate and Exchange Purchase.
    - Probability 400+% FPL uninsured → 3.6%

# Results: Demand Estimation

	(1)	(2)	(3)	(4)	(5)
	Log Share of HIX 100-150% FPL	Premiums (000s):	P(HIX Purchase):	Unsubsidized Uninsured Rate (% point):	Demand IV ((3)/(2))
Estimated ATT	-0.471***	-0.374***	0.029***	-0.126*	-0.078***
90% Confidence Interval	[-0.511, -0.412]	[-0.511, -0.213]	[0.012, 0.043]	[-0.281, -0.02]	[-0.125, -0.05]
Implied Intercept	-	-	-	-	0.981***
90% Confidence Interval					[0.887, 1.14]
Person Controls?	No	No	Yes	Yes	-
Plan Controls?	No	Yes	No	No	-
Year Fixed Effects?	Yes	Yes	Yes	Yes	-
Rating-Area Fixed Effects?	Yes	Yes	No	No	-
State Fixed Effects?	No	No	Yes	Yes	-
Pre-Expansion Sample Mean	0.415	3.09	0.74	3.6	-
Implied Post-Expansion Mean	0.259	2.71	0.769	3.5	-
Implied No-MHIU Mean	0	2.09	0.818	3.3	-
Observations	747	59,013	213,208	3,595,818	-

Placebo Check

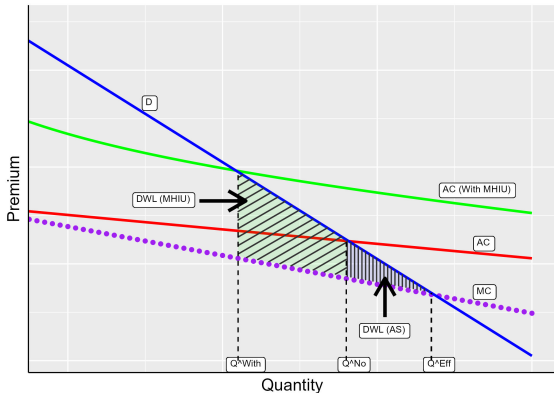
# Welfare: Setup



# Theory Reminder:

We are estimating:

- Size of MHIU DWL
- Size of AS DWL
- Enrollment change for AS
- Enrollment change for MHIU



# Welfare Estimates

	(1)	(2)	(3)	(4)	(5)	(6)
	Enrollment loss due to MHIU	Enrollment loss due to MHIU and AS	Welfare loss ( $\Delta_{MFE}$ )	Loss due to AS ( $\Delta_{AGE}$ )	Loss due to MHIU ( $\Delta_{MFGA}$ )	Share of Welfare Loss due to MHIU
1 RAND Elasticity (Base Case)	0.028	0.2	177	132	46	0.257

- Base Case: \$50 DWL from MHIU
  - ▶ Compared to \$130 from AS.
- Enrollment loss from AS is much bigger.
  - ▶ But welfare losses are still comparable.

# Welfare Estimates

	(1)	(2)	(3)	(4)	(5)	(6)
	Enrollment loss due to MHIU	Enrollment loss due to MHIU and AS	Welfare loss ( $\Delta_{MFE}$ )	Loss due to AS ( $\Delta_{AGE}$ )	Loss due to MHIU ( $\Delta_{MFGA}$ )	Share of Welfare Loss due to MHIU
1 RAND Elasticity (Base Case)	0.028	0.2	177	132	46	0.257
2 Lavetti et al Elasticity	0.018	0.188	157	128	28	0.182
3 Ellis et al Elasticity	0.061	0.241	256	143	113	0.441
4 Brot-Goldberg et al Elasticity	0.079	0.241	280	132	148	0.529

- Changing the elasticity ( $\mu$ ) assumption has a big effect.
  - ▶ Lavetti et al is a lower bound and drops the DWL to \$28.

# Welfare Estimates

		(1)	(2)	(3)	(4)	(5)	(6)
		Enrollment loss due to MHIU	Enrollment loss due to MHIU and AS	Welfare loss ( $\Delta_{MFE}$ )	Loss due to AS ( $\Delta_{AGE}$ )	Loss due to MHIU ( $\Delta_{MFGA}$ )	Share of Welfare Loss due to MHIU
1	RAND Elasticity (Base Case)	0.028	0.2	177	132	46	0.257
5	No Adverse Selection	0.023	0.023	3	0	3	1
6	1/2 as Much Adverse Selection	0.025	0.095	49	27	23	0.46
7	2x More Adverse Selection	0.034	0.241	390	319	72	0.184

- Changing the level of AS has a huge effect.
- No adverse selection reduces the DWL to basically 0.
- Increasing adverse selection also increases the loss from MHIU.



# Welfare Estimates

	(1)	(2)	(3)	(4)	(5)	(6)
	Enrollment loss due to MHIU	Enrollment loss due to MHIU and AS	Welfare loss ( $\Delta_{MFE}$ )	Loss due to AS ( $\Delta_{AGE}$ )	Loss due to MHIU ( $\Delta_{MFGA}$ )	Share of Welfare Loss due to MHIU
1 RAND Elasticity (Base Case)	0.028	0.2	177	132	46	0.257
8 If markups are 15%	0.034	0.241	269	201	68	0.254
9 Health/Wealth Gradient	0.028	0.241	263	207	56	0.214

- Allowing for 15% markups also increases the welfare loss.
- Adding a \$1000 health/wealth gradient does as well.
  - ▶ Since subsidies are means-tested, it basically just mimics MHIU

## Conclusion:

- We show how community rating with means-tested OOP subsidies can unravel a market through moral hazard.
- Use the ACA HIXs and Medicaid expansion to test this empirically.
- Economically meaningful:
  - ▶ Accounts for 12% higher premiums.
  - ▶ Total welfare loss of \$177 per person – about a quarter due to MHIU.

Thank you! Questions?

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# Results: Placebo Check

	150-400% FPL:	400+% FPL:
	(1)	(2)
	P(HIX Purchase):	Employer-Sponsored Insurance Rate:
Estimated ATT	-0.001	0.001
90% Confidence Interval	[-0.014, 0.013]	[-0.001, 0.004]
Pre-Expansion Sample Mean	0.435	0.962
Observations	499,980	3,595,818

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