

# Policy Evaluation - Natural Experiments & the Oregon Medicaid Study

March 19 & 21, 2023

## Estimating Treatment Effects Review

- $ATE = Avg_n[Y_i^1 - Y_i^0]$
- $ATE_{est} = Avg_n[Y_i^1 | D_i = 1] - Avg_n[Y_i^0 | D_i = 0]$
- When  $(Y^1, Y^0) \not\perp D$ :

$$ATE_{est} = ATE + \underbrace{\{Avg_n[Y_i^0 | D_i = 1] - Avg_n[Y_i^0 | D_i = 0]\}}_{\text{Selection Bias}} \\ + \underbrace{(1 - \pi)(ATT - ATU)}_{\text{Heterogeneous Treatment Effect Bias}}$$

- $ATE_{est} = \beta_0 + \beta_1 D + \beta_2 X_1 + \beta_3 X_2 + \dots \beta_k X_{k-1} + \varepsilon$

# Natural Experiments to Estimate Treatment Effects

- What is a natural experiment?

# Natural Experiments to Estimate Treatment Effects



VS.



**EXPERIMENTAL GROUP**

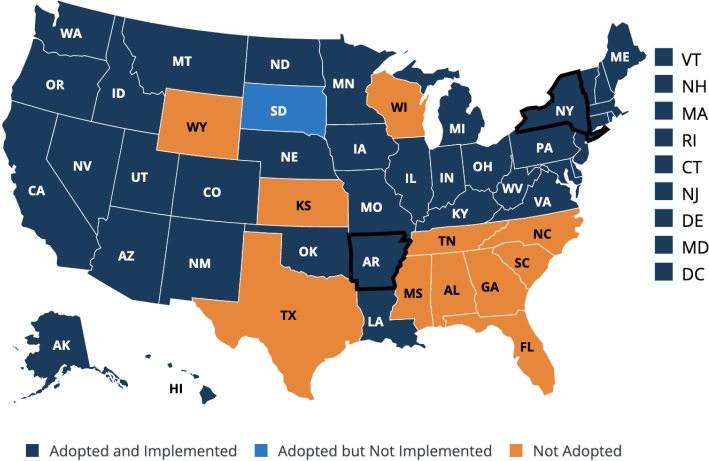


**CONTROL GROUP**



# Natural Experiments to Estimate Treatment Effects

## Status of State Action on the Medicaid Expansion Decision



# Natural Experiments to Estimate Treatment Effects

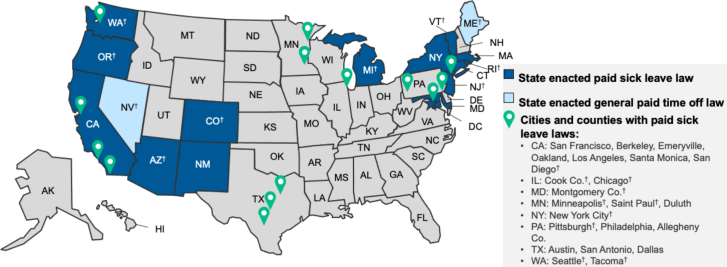


SOURCE: U.S. Geological Survey photo, May 2002

JOHN DUCHNESKIE / Inquirer Staff Artist

# Natural Experiments to Estimate Treatment Effects

## State and Local Paid Sick Leave Laws, 2021



†Law permits use of accrued leave for workplace closure or closure of the worker's child's school or childcare associated with a public health emergency.  
NOTES: NM's law takes effect July 1, 2022. CO's law for employers with fewer than 16 workers takes effect Jan. 1, 2022; currently in effect for all other CO employers. Allegheny Co.'s law was enacted in Sept. 2021 and will take effect 90 days after the county posts compliance information for employers. The three local laws passed in TX are on hold due to a pending court challenge. All other state and local laws are currently in effect. All state and all local paid sick leave laws except Pittsburgh, Oakland, and Berkeley permit use of paid leave for reasons associated with sexual assault, domestic violence, or stalking, known as "safe time."  
SOURCE: KFF analysis of state paid family and medical leave laws; A Better Balance. [Overview of Paid Sick Time laws in the United States](#)



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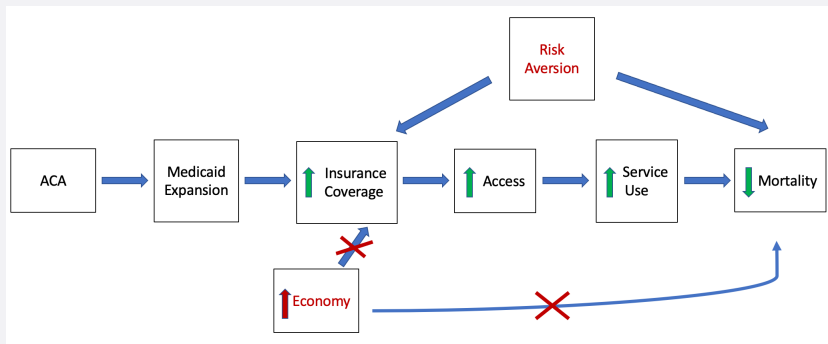
SPECIAL ARTICLE

### Cancer Screening after the Adoption of Paid-Sick-Leave Mandates

Kevin Callison, Ph.D., Michael F. Pesko, Ph.D., Serena Phillips, Dr.P.H., and Julie A. Sosa, M.D.

## Natural Experiments to Estimate Treatment Effects

- Why are natural experiments valuable when estimating treatment effects?





# Oregon Health Study

- Oregon Medicaid

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- Oregon Medicaid
  - ▶ Oregon Health Plan Plus (OHP Plus) - coverage for categorically eligible.
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- OHP Standard enrollment:
  - ▶ 2002: 110k
  - ▶ 2004: Closed to new enrollment
  - ▶ 2008: 19k

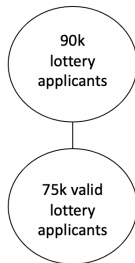
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- OHP Standard enrollment:
  - ▶ 2002: 110k
  - ▶ 2004: Closed to new enrollment
  - ▶ 2008: 19k
- Expand OHP Standard enrollment by 10k in 2008.
  - ▶ 90k people applied

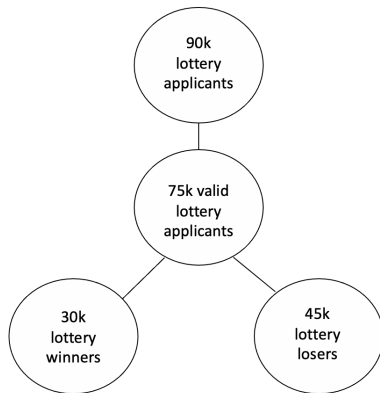
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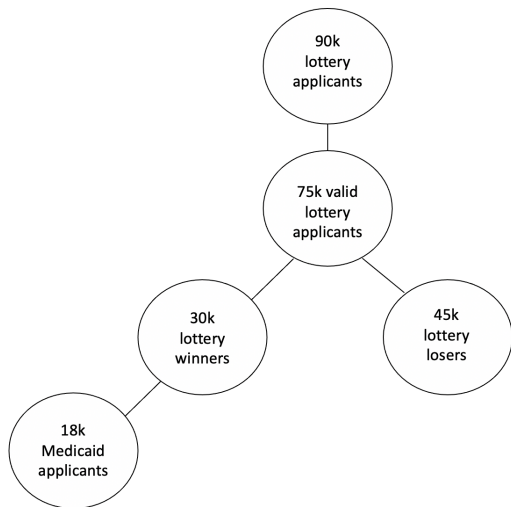
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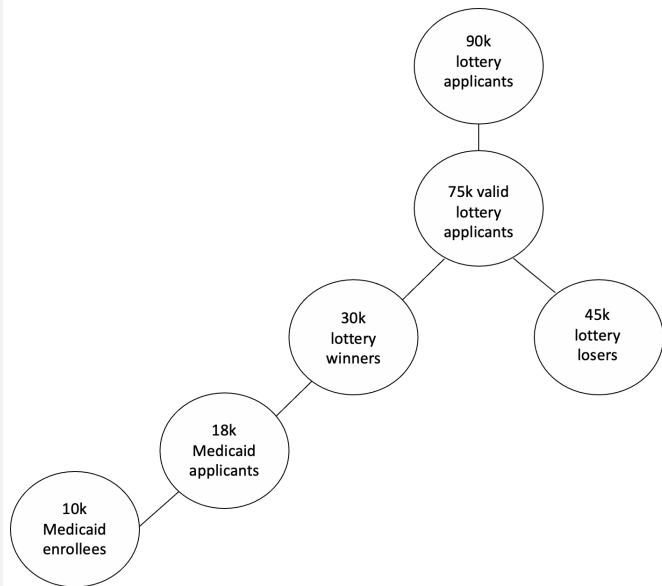


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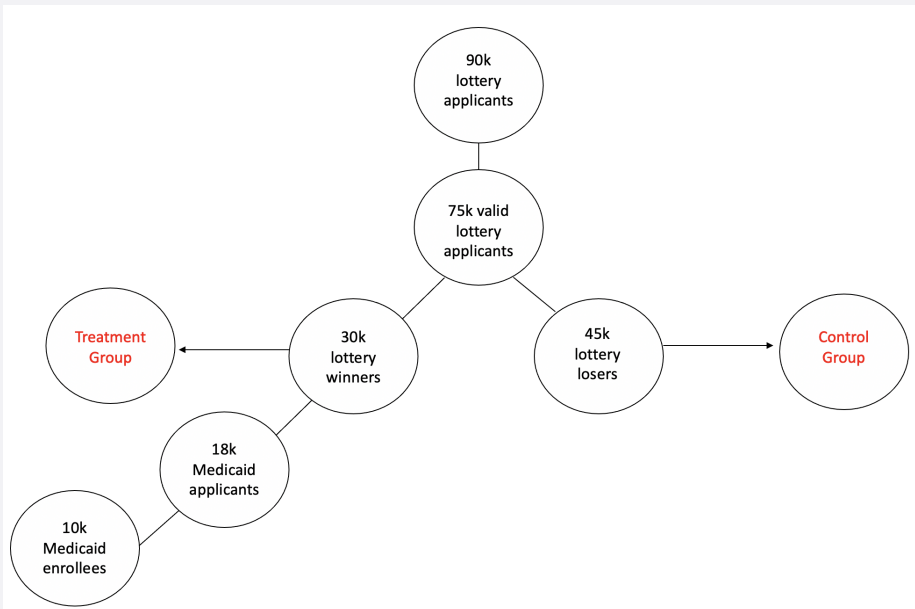




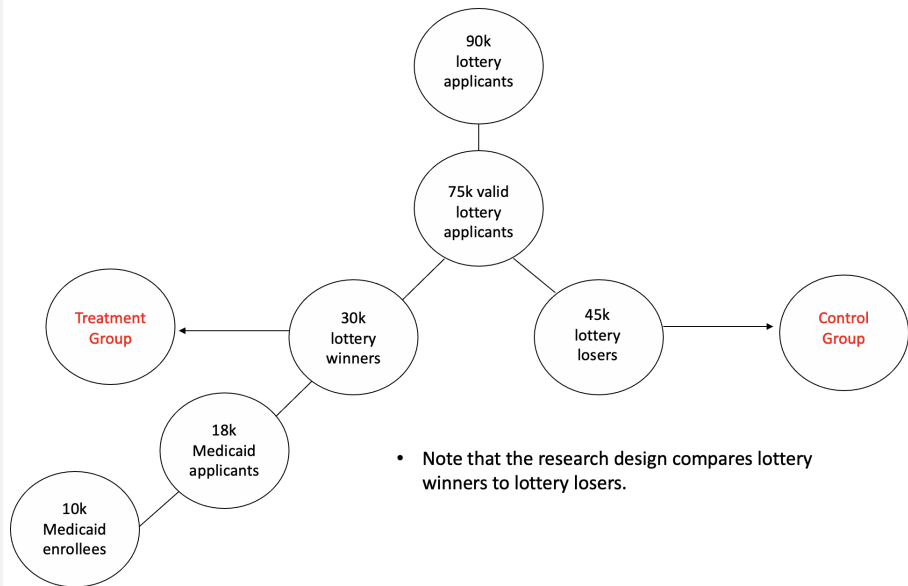
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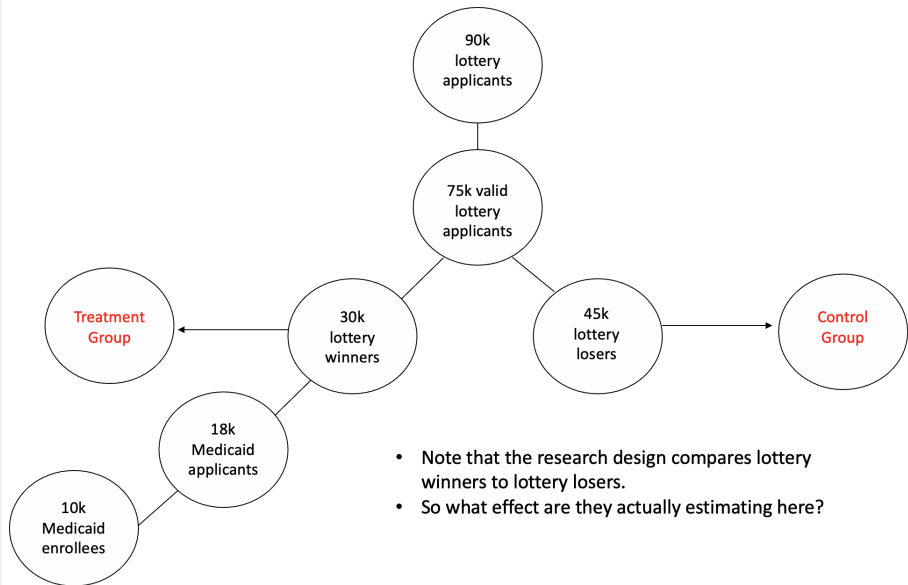
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- Note that the research design compares lottery winners to lottery losers.
- So what effect are they actually estimating here?

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  2. Instrumental variables estimate of the local average treatment effect (LATE).

## Oregon Health Study

- **Instrumental Variables - Two-stage least squares**
- First stage:  $Insurance_i = \alpha_0 + \alpha_1 Lottery_i + \varepsilon_i$
- Second stage:  $Health_i = \beta_0 + \beta_1 \hat{Insurance}_i + \varepsilon_i$

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- ▶ LATE: those who obtain insurance after winning the lottery and who would not have obtained insurance without winning the lottery.
  - How does this distinction between ITT and TOT/LATE relate to the concept of external validity?

- Balance Test

**Table A13: Lottery list characteristics and balance of treatment and controls**

	Control Mean (std dev)	Difference between treatment and control		
		Full sample	Credit report subsample	Survey respondents subsample
	(1)	(2)	(3)	(4)
<b>Panel A: Lottery list characteristics</b>				
Year of birth	1968.0 (12.255)	0.162 (0.100)	0.136 (0.119)	-0.066 (0.191)
Female	0.557 (0.497)	-0.0069 (0.0033)	-0.0027 (0.0041)	-0.0042 (0.0068)
English as preferred language	0.922 (0.268)	0.0024 (0.0026)	0.0042 (0.0029)	-0.00033 (0.0048)
Signed up self	0.918 (0.274)	0.00030 (0.00028)	0.00060 (0.0010)	-0.0016 (0.0027)
Signed up first day of lottery	0.093 (0.290)	0.0012 (0.0025)	0.00093 (0.0031)	0.0061 (0.0049)
Gave phone number	0.862 (0.345)	-0.0029 (0.0028)	0.000088 (0.0034)	0.0059 (0.0037)
Address a PO Box	0.117 (0.321)	0.00044 (0.0027)	0.0023 (0.0034)	-0.0023 (0.0053)
In MSA	0.773 (0.419)	-0.0024 (0.0036)	-0.0018 (0.0044)	0.0011 (0.0070)
Zip code median household income	39,265.4 (8463.542)	44.891 (72.887)	12.998 (89.653)	22.031 (135.815)

- Results - Hospital Admissions

## HOSPITAL UTILIZATION

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel A: Extensive margin				
All hospital admissions	0.067 (0.250)	0.0054 (0.0019)	0.021 (0.0074)	[0.004]
Admissions through ER	0.048 (0.214)	0.0018 (0.0016)	0.0070 (0.0062)	[0.265]
Admissions not through ER	0.029 (0.167)	0.0041 (0.0013)	0.016 (0.0051)	[0.002]

- Results - Hospital Admissions

HOSPITAL UTILIZATION				
	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
<b>Panel B: All hospital admissions</b>				
Days	0.498 (3.795)	0.026 (0.027)	0.101 (0.104)	[0.329] {0.328}
List charges	2,613 (19,942)	258 (146)	1,009 (569)	[0.077] {0.106}
Procedures	0.155 (1.08)	0.018 (0.0083)	0.070 (0.032)	[0.031] {0.059}

- Results - Hospital Admissions

HOSPITAL UTILIZATION				
	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel C: Admissions through ER				
Days	0.299 (2.326)	0.023 (0.017)	0.089 (0.067)	[0.183] {0.187}
List charges	1,502 (12,749)	163 (96)	636 (376)	[0.091] {0.171}
Procedures	0.081 (0.694)	0.0080 (0.0054)	0.031 (0.021)	[0.135] {0.187}

- Results - Hospital Admissions

HOSPITAL UTILIZATION				
	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel D: Admissions not through ER				
Days	0.199 (2.38)	0.0033 (0.017)	0.013 (0.065)	[0.841] {0.842}
List charges	1,110 (12,422)	98 (91)	384 (356)	[0.281] {0.383}
Procedures	0.075 (0.708)	0.010 (0.0056)	0.038 (0.022)	[0.080] {0.162}

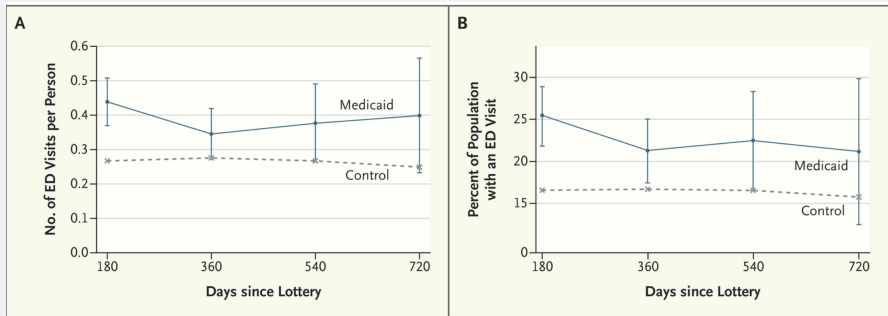


## ● Results - ED Visits

### Perspective

## Effect of Medicaid Coverage on ED Use — Further Evidence from Oregon's Experiment

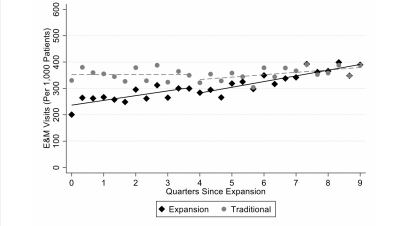
Amy N. Finkelstein, Ph.D., Sarah L. Taubman, Ph.D., Heidi L. Allen, Ph.D., Bill J. Wright, Ph.D., and Katherine Baicker, Ph.D.



# Louisiana Medicaid Expansion

- Results - E&M and ED Visits

Figure 1: Evaluation and Management Visits for Expansion vs. Non-Expansion Beneficiaries



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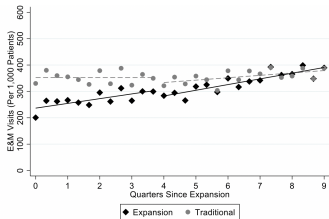
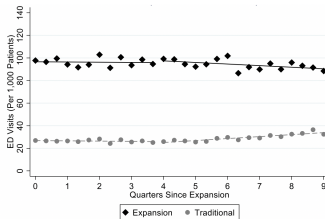


Figure 2: Emergency Department Visits for Expansion vs. Non-Expansion Beneficiaries



- Results - Preventive Care

COMPLIANCE WITH RECOMMENDED PREVENTIVE CARE (SURVEY DATA)

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Blood cholesterol checked (ever)	0.625 (0.484)	0.033 (0.0074)	0.114 (0.026)	[<0.0001] [<0.0001]
Blood tested for high blood sugar/diabetes (ever)	0.604 (0.489)	0.026 (0.0074)	0.090 (0.026)	[0.0004] [<0.0001]
Mammogram within last 12 months (women $\geq$ 40)	0.298 (0.457)	0.055 (0.012)	0.187 (0.04)	[<0.0001] [<0.0001]
Pap test within last 12 months (women)	0.406 (0.491)	0.051 (0.01)	0.183 (0.034)	[<0.0001] [<0.0001]

- Results - Financial Strain

**FINANCIAL STRAIN (ADMINISTRATIVE DATA)**

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel A: Overall				
Any bankruptcy	0.014 (0.119)	0.0022 (0.0014)	0.0086 (0.0053)	[0.106] {0.358}
Any lien	0.021 (0.144)	0.0012 (0.0014)	0.0047 (0.0056)	[0.406] {0.698}
Any judgment	0.064 (0.244)	0.0014 (0.0024)	0.0054 (0.010)	[0.573] {0.698}
Any collection	0.500 (0.500)	-0.012 (0.0041)	-0.048 (0.016)	[0.003] {0.013}
Any delinquency (credit accounts)	0.366 (0.482)	0.0016 (0.0042)	0.0063 (0.017)	[0.704] {0.698}

- Results - Financial Strain

**FINANCIAL STRAIN (ADMINISTRATIVE DATA)**

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
<b>Panel B: Medical debt</b>				
Any medical collection	0.281 (0.449)	-0.016 (0.0040)	-0.064 (0.016)	[<0.0001] {<0.0001}
Amount owed in medical collections	1,999 (6733)	-99 (45)	-390 (177)	[0.028] {0.025}

- Results - Financial Strain

## FINANCIAL STRAIN (ADMINISTRATIVE DATA)

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel C: Nonmedical debt				
Any nonmedical collection	0.392 (0.488)	-0.0046 (0.0041)	-0.018 (0.016)	[0.264] {0.455}
Amount owed in nonmedical collections	2,740 (9,492)	-20 (63)	-79 (248)	[0.751] {0.752}

- Results - Financial Strain

## FINANCIAL STRAIN (SURVEY DATA)

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Any out of pocket medical expenses, last six months	0.555 (0.497)	-0.058 (0.0077)	-0.200 (0.026)	[<0.0001] {<0.0001}
Owe money for medical expenses currently	0.597 (0.491)	-0.052 (0.0076)	-0.180 (0.026)	[<0.0001] {<0.0001}
Borrowed money or skipped other bills to pay medical bills, last six months	0.364 (0.481)	-0.045 (0.0073)	-0.154 (0.025)	[<0.0001] {<0.0001}
Refused treatment because of med- ical debt, last six months	0.081 (0.273)	-0.011 (0.0041)	-0.036 (0.014)	[0.01] {0.01}



- Results - Health

HEALTH				
	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel A: Administrative data				
Alive	0.992 (0.092)	0.00032 (0.00068)	0.0013 (0.0027)	[0.638]
Panel B: Survey data				
Self-reported health good/very good/excellent (not fair or poor)	0.548 (0.498)	0.039 (0.0076)	0.133 (0.026)	[<0.0001] [<0.0001]
Self-reported health not poor (fair, good, very good, or excellent)	0.86 (0.347)	0.029 (0.0051)	0.099 (0.018)	[<0.0001] [<0.0001]
Health about the same or gotten better over last six months	0.714 (0.452)	0.033 (0.0067)	0.113 (0.023)	[<0.0001] [<0.0001]
# of days physical health good, past 30 days*	21.862 (10.384)	0.381 (0.162)	1.317 (0.563)	[0.019] [0.018]
# days poor physical or mental health did not impair usual activity, past 30 days*	20.329 (10.939)	0.459 (0.175)	1.585 (0.606)	[0.009] [0.015]
# of days mental health good, past 30 days*	18.738 (11.445)	0.603 (0.184)	2.082 (0.64)	[0.001] [0.003]
Did not screen positive for depression, last two weeks	0.671 (0.470)	0.023 (0.0071)	0.078 (0.025)	[0.001] [0.003]

## ● Results - Mechanisms

POTENTIAL MECHANISMS FOR IMPROVED HEALTH (SURVEY DATA)

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel A: Access to care				
Have usual place of clinic-based care	0.499 (0.500)	0.099 (0.0080)	0.339 (0.027)	[<0.0001]
Have personal doctor	0.490 (0.500)	0.081 (0.0077)	0.280 (0.026)	[<0.0001]
Got all needed medical care, last six months	0.684 (0.465)	0.069 (0.0063)	0.239 (0.022)	[<0.0001]
Got all needed drugs, last six months	0.765 (0.424)	0.056 (0.0055)	0.195 (0.019)	[<0.0001]
Didn't use ER for nonemergency, last six months	0.916 (0.278)	-0.0011 (0.0043)	-0.0037 (0.015)	[0.804]

- Results - Mechanisms

POTENTIAL MECHANISMS FOR IMPROVED HEALTH (SURVEY DATA)

	Control mean (1)	ITT (2)	LATE (3)	<i>p</i> -values (4)
Panel B: Quality of care				
Quality of care received last six months good/very good/excellent (conditional on any)	0.708 (0.455)	0.043 (0.0081)	0.142 (0.027)	[<0.0001]
Panel C: Happiness				
Very happy or pretty happy (vs. not too happy)	0.594 (0.491)	0.056 (0.0074)	0.191 (0.026)	[<0.0001]

## ● Results - Biometrics

Variable	Mean Value in Control Group	Change with Medicaid Coverage (95% CI) <sup>†</sup>	P Value
<b>Blood pressure</b>			
Systolic (mm Hg)	119.3±16.9	-0.52 (-2.97 to 1.93)	0.68
Diastolic (mm Hg)	76.0±12.1	-0.81 (-2.65 to 1.04)	0.39
Elevated (%) <sup>‡</sup>	16.3	-1.33 (-7.16 to 4.49)	0.65
<b>Hypertension</b>			
Diagnosis after lottery (%) <sup>§¶</sup>	5.6	1.76 (-1.89 to 5.40)	0.34
Current use of medication for hypertension (%) <sup>§  </sup>	13.9	0.66 (-4.48 to 5.80)	0.80
<b>Cholesterol<sup>***</sup></b>			
Total level (mg/dl)	204.1±34.0	2.20 (-3.44 to 7.84)	0.45
High total level (%)	14.1	-2.43 (-7.75 to 2.89)	0.37
HDL level (mg/dl)	47.6±13.1	0.83 (-1.31 to 2.98)	0.45
Low HDL level (%)	28.0	-2.82 (-10.28 to 4.64)	0.46
<b>Hypercholesterolemia</b>			
Diagnosis after lottery (%) <sup>§¶</sup>	6.1	2.39 (-1.52 to 6.29)	0.23
Current use of medication for high cholesterol level (%) <sup>§  </sup>	8.5	3.80 (-0.75 to 8.35)	0.10
<b>Glycated hemoglobin</b>			
Level (%)	5.3±0.6	0.01 (-0.09 to 0.11)	0.82
Level ≥6.5% (%) <sup>††</sup>	5.1	-0.93 (-4.44 to 2.59)	0.61
<b>Diabetes</b>			
Diagnosis after lottery (%) <sup>§¶</sup>	1.1	3.83 (1.93 to 5.73)	<0.001
Current use of medication for diabetes (%) <sup>§  </sup>	6.4	5.43 (1.39 to 9.48)	0.008

## ● Results - Biometrics

Variable	Mean Value in Control Group	Change with Medicaid Coverage (95% CI) <sup>†</sup>	P Value
Depression			
Positive screening result (%) <sup>‡‡</sup>	30.0	-9.15 (-16.70 to -1.60)	0.02
Diagnosis after lottery (%) <sup>¶¶</sup>	4.8	3.81 (0.15 to 7.46)	0.04
Current use of medication for depression (%) <sup>§§</sup>	16.8	5.49 (-0.46 to 11.45)	0.07
Framingham risk score (%) <sup>§§§</sup>			
Overall	8.2±7.5	-0.21 (-1.56 to 1.15)	0.76
High-risk diagnosis	11.6±8.3	1.63 (-1.11 to 4.37)	0.24
Age of 50–64 yr	13.9±8.2	-0.37 (-2.64 to 1.90)	0.75

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- ▶ Biometrics

- Elevated blood pressure ↓ 0.8% (p=0.65)
- High cholesterol ↓ 17% (p=0.37)
- Framingham high-risk ↑ 14% (p=0.24)
- Depression medication ↑ 33% (p=0.07)
- Positive depression screening ↓ 31% (p=0.02)
- Mortality ↓ 0.14% (p=0.638)

- Results Summary - Survey Data:
  - ▶ Preventive Care
    - Cholesterol check ↑ 18% ( $p < 0.001$ )
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- ▶ Mechanisms

- Usual source of care ↑ 68% ( $p < 0.001$ )
- Got all needed care ↑ 35% ( $p < 0.001$ )



## Oregon Health Study

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- The Oregon Medicaid Study found no statistically significant effects of winning the lottery on biometric measures of health. Frakt argues that we should not interpret these findings as evidence that Medicaid does not improve health. Why not?
- What is the "Peltzman Effect" and how does it apply to the findings from the Oregon Health Study?

## Randomization vs. Regression

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- How important is selection bias in this case?
- *“Results from the randomized evaluation showed that Medicaid substantially improved self-reported health. However if we analyzed the same data using observational methods rather than taking advantage of the randomization, results suggested that Medicaid actually worsens these same self-reported health measures.”*

- How important is selection bias in this case?

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### ORIGINAL ARTICLES

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## Primary Payer Status Affects Mortality for Major Surgical Operations

*Damien J. LaPar, MD,\* Castigliano M. Bhamidipati, DO,\* Carlos M. Mery, MD, MPH,\*  
George J. Stukenborg, PhD,† David R. Jones, MD,\* Bruce D. Schirmer, MD,\* Irving L. Kron, MD,\*  
and Gorav Ailawadi, MD\**

Multivariable logistic regression was performed to calculate the adjusted odds of in-hospital death and in-hospital complications among patients undergoing major surgical operations. All preoperative variables entered as covariates (patient age, gender, elective operative status, mean income, hospital geographic region, teaching hospital status, type of operation, primary payer status, and categories for comorbid disease) were selected a priori based upon established clinical risk or were considered potential confounders for the effect of payer status among patients. All covariates contributing cases to each estimated outcome,

## Randomization vs. Regression

	Private	Medicare	Uninsured	Medicaid
Mortality	1.00	1.54	1.74	1.97
Length of Stay	7.38	8.77	7.01	10.49
Total Costs	63,057	69,408	65,667	79,140

# Medicaid Is Worse Than No Coverage at All

New research shows that patients on this government plan fare poorly. So why does the president want to shove one in four Americans into it?

Mar 2, 2011, 12:23pm EST

By Scott Gottlieb

Updated March 10, 2011 12:01 am ET

## Medicaid's awful results: Column

Glenn Harlan Reynolds

Published 11:48 a.m. ET Nov. 11, 2013

CRITICAL CONDITION

## Why Medicaid is a Humanitarian Catastrophe



Avik Roy Forbes Staff

The Apothecary Contributor Group

Policy

Commentary from Forbes' Policy Editor

## UVa Study: Surgical Patients on Medicaid Are 13% More Likely to Die Than Those Without Insurance

By AVIK ROY | July 17, 2010 11:36 PM

