

# Health Screening and Selection: Evidence from Biennial Subsidies in South Korea

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# Health screening is an important preventive care

- Health screenings are medical tests to find diseases early
  - Cancer (20%)
  - Cardiovascular (22%) + diabetes (3%) + stroke (5%)
- Public support for screenings
  - US spending on 5 cancer screenings: \$43 billion / year ( $\approx \$130/\text{person}$ )  
[\(Halpern et al., 2024\)](#)
  - ACA preventive care mandate
  - Government-run public screening programs

## We do not want to screen everyone

- Costs and benefits of screening
  - Benefits: early cancer diagnosis ↑, mortality ↓
  - Costs: false positives, overdiagnosis, and unnecessary treatment  
([Welch and Passow, 2014](#); [Welch et al., 2016](#); [Autier et al., 2017](#))

# We do not want to screen everyone

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([Welch and Passow, 2014](#); [Welch et al., 2016](#); [Autier et al., 2017](#))
- We want to screen **high-risk** individuals
  - (Benefits > Costs): Unhealthy (high-risk) people more likely to have undiagnosed conditions
  - (Costs > Benefits): Healthy (low-risk) people without any disease to find ([Kowalski, 2023](#))

## High-risk individuals are less likely to participate in public screenings

- Who participates in screening?
  - High-risk: family history or environmental risk exposure
  - Low-risk: attentiveness to health or better access to health care

# High-risk individuals are less likely to participate in public screenings

- Who participates in screening?
    - High-risk: family history or environmental risk exposure
    - **Low-risk**: attentiveness to health or better access to health care
  
  - In **public** health screenings, participants show
    - Higher socioeconomic status
    - Better health behaviors
    - Lower mortality
- (Pill et al., 1988; Waller et al., 1990; Bender et al., 2015; Carethers and Doubeni, 2020)

## How can we target health screenings better?

If health screenings are subsidized, who are the marginal participants?

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- Healthier (low-risk) individuals
  - Moral hazard
  - Rationale for cost-sharing

(Arrow, 1963; Pauly, 1968; Zeckhauser, 1970; Newhouse, 1993; Manning and Marquis, 1996)

# How can we target health screenings better?

If health screenings are subsidized, who are the marginal participants?

- Healthier (low-risk) individuals
  - Moral hazard
  - Rationale for cost-sharing

(Arrow, 1963; Pauly, 1968; Zeckhauser, 1970; Newhouse, 1993; Manning and Marquis, 1996)

- Less healthy (high-risk) individuals
  - Financial constraints
  - Behavioral hazard (present bias, salience)

(Baicker et al., 2015; Choudhry et al., 2011; Bai et al., 2021)

## Research setting and questions

- National Health Screening Program in South Korea
  - Nationwide program for all the citizens
  - Blanket subsidies at even-numbered ages
  - Repeated throughout lifetime

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- Research questions
  1. Who are most responsive to the subsidies?
  2. What are the causal effects of screening on disease diagnosis and preventive care use?

# Preview of results

## 1. Strong first stage

- Take-up increases by 16-19%P
- Positive spillovers to unsubsidized screenings

## 2. High-risk compliers

- Compliers are from lower socioeconomic backgrounds with poorer health conditions

## 3. Cancer diagnosis & preventive care use ↑

- Cancer diagnoses increase by 17-19%
- Treatment for cancer precursors and cardiovascular risk factors increases

# Contributions to health screening literature

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Literature	Prior studies	Gap	Contribution
Selection into screening <sup>1</sup>			

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Causal effects of screening	Limited type of screening & aggregate diagnosis	Impact of other screenings & early diagnosis	7 cancer screenings & in-situ diagnosis

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# National Health Screening Program

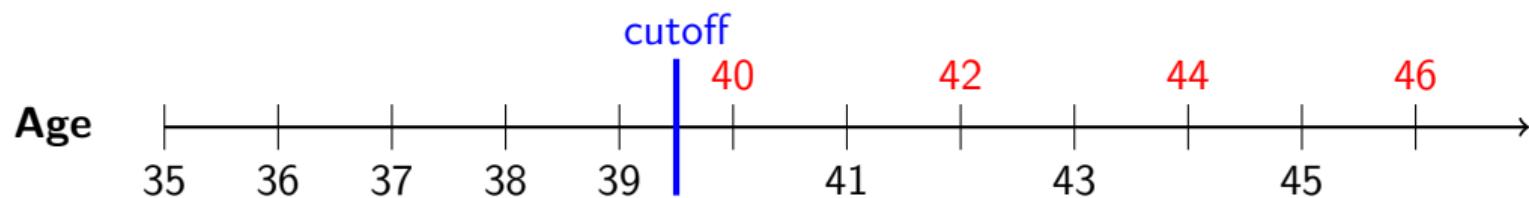
- General health screening

- Height, weight, waist circumference
- Sight, hearing
- Blood pressure, blood sugar and cholesterol level
- Chest X-ray (tuberculosis), bone density (osteoporosis)

- Cancer screenings

Screening	Tests
Stomach	Upper endoscopy
Breast	Mammography
Cervical	Pap smear
Liver	Alpha-fetoprotein blood test & sonogram
Colorectal	Fecal occult blood test → colonoscopy

## Biennial health screening subsidies



- From age 40
- Biennial subsidies at even ages
- Age = Current year - Birth year
- Coordination following clinical guidelines

# Subsidy schedule across screenings

	Biennial subsidy				Annual subsidy		No subsidy	
	General	Stomach	Breast	Cervical	Liver	Colorectal	Lung	Prostate
Age cutoff	40	40	40	30	40	50		
Subsidy amount	100%	90%	90%	100%	90%	90%	0%	0%
Full cost (\$)	50	65	40	15	90	10	100	20
Target		Female	Female	High risk group			Male	

⇒ Total eligible subsidy amount: \$115 (men), \$165 (women)

⇒ 4.5-6.5% of annual health care expenditure / person

Liver high risk group

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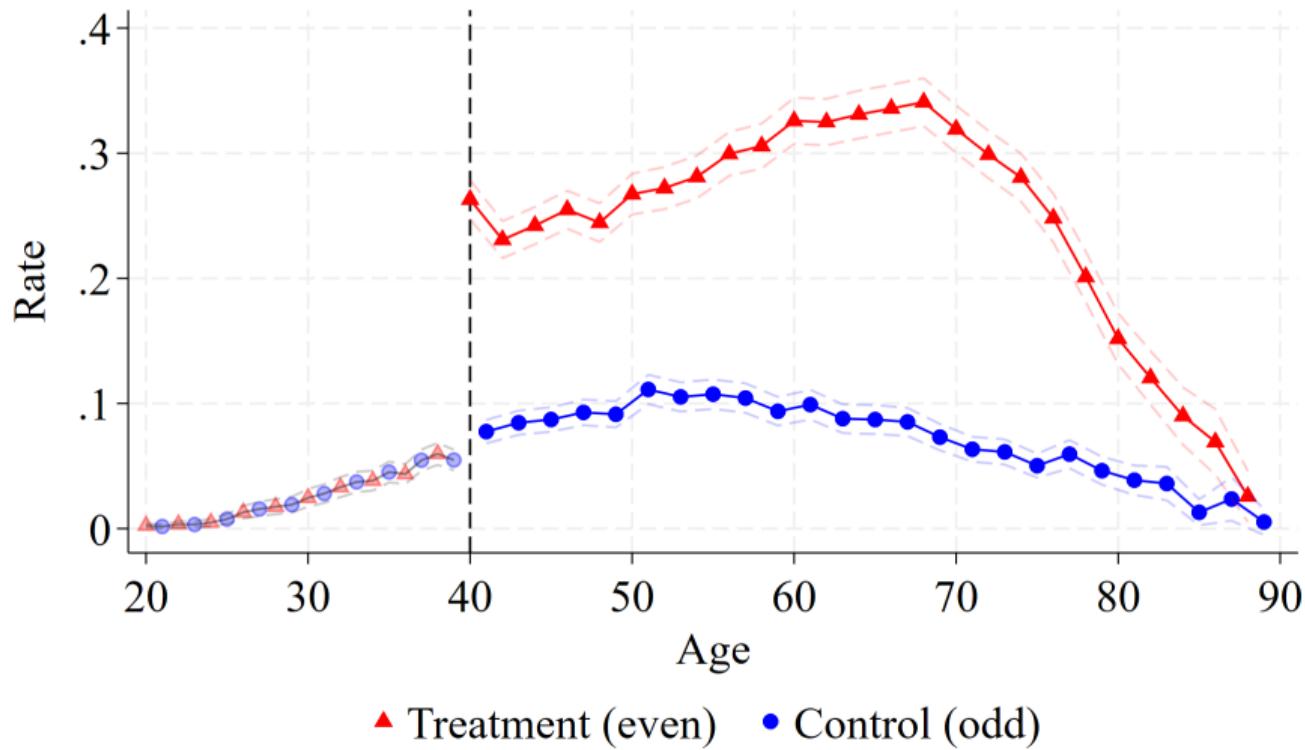
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## Stomach cancer screening take-up by age



## Comparing even with odd age groups from age 40

- Even- and odd-aged individuals are balanced conditional on  $f(\text{age})$
- Econometric specification

$$\text{screen}_{it} = \beta_0 + \beta_1 \cdot \text{treat}_{it} + f(\text{age}_{it}) + \epsilon_{it} \quad (1)$$

- Individual  $i$  in year  $t$
- Analytical sample:  $\text{age} \in [40, 89]$
- $\text{treat}_{it} = 1$  if individual  $i$  turns even-numbered age in year  $t$
- $f(\text{age})$ : linear splines with 5 years interval
- Standard error clustered at the individual level

# Treatment and control groups are balanced conditional on $f(\text{age})$

	(1)	(2)	(3)
	Treatment (Even)	Control (Odd)	Conditional difference
Age	58.697 (12.532)	59.240 (12.353)	- -
Female	0.530 (0.499)	0.532 (0.499)	-0.002* (0.001)
Currently married	0.799 (0.401)	0.798 (0.402)	-0.0011 (0.0008)
Years of education	10.320 (4.510)	10.227 (4.538)	-0.003 (0.008)
Working status	0.610 (0.488)	0.608 (0.488)	-0.003* (0.001)
Individual income	1446.3 (2081.6)	1425.7 (2068.1)	2.762 (5.185)
Household income	4104.4 (3708.6)	4086.7 (3737.9)	3.221 (14.267)
Own a house	0.734 (0.442)	0.737 (0.441)	-0.0002 (0.0011)
Number of household members	3.067 (1.317)	3.051 (1.317)	-0.004 (0.003)
N	54,274	52,909	
Share	(0.51)	(0.49)	
F(8, 15939)			1.65
Prob > F			(0.10)

## Identifying variation with intertemporal substitution

- Identifying variation: short-run temporal variation in subsidy eligibility

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- Two sources of identifying variation

Even/Odd take-up gap = Net increase + Intertemporal substitution

net increase

substitution

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Even/Odd take-up gap = Net increase + Intertemporal substitution

net increase

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- Interpretation of treatment effects
  - ⇒ **Short-run** effect of subsidy eligibility **this year**, generated by net increase in screening take-up and shift in screening timing

# Survey data

- Korean Health Panel Survey dataset (2008-2018)
  - Analytical sample: 107,200 individual-year
  - Demographics / health care utilization / health behaviors
- Health screening take-up well recorded
  - Date, screening type, tests performed, screening results

Data collection

## Administrative data

- National health insurance claims data (2002-2021)
  - Standard cohort: 8,670,000 individual-year
  - Customized cohort: 7,450,000 individual-year
- Cancer diagnoses available
  - Coinsurance reduction program for rare and severe diseases
  - True positive cancer diagnoses
  - Including cancers diagnosed without a screening
- Incomplete health screening records

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## Results

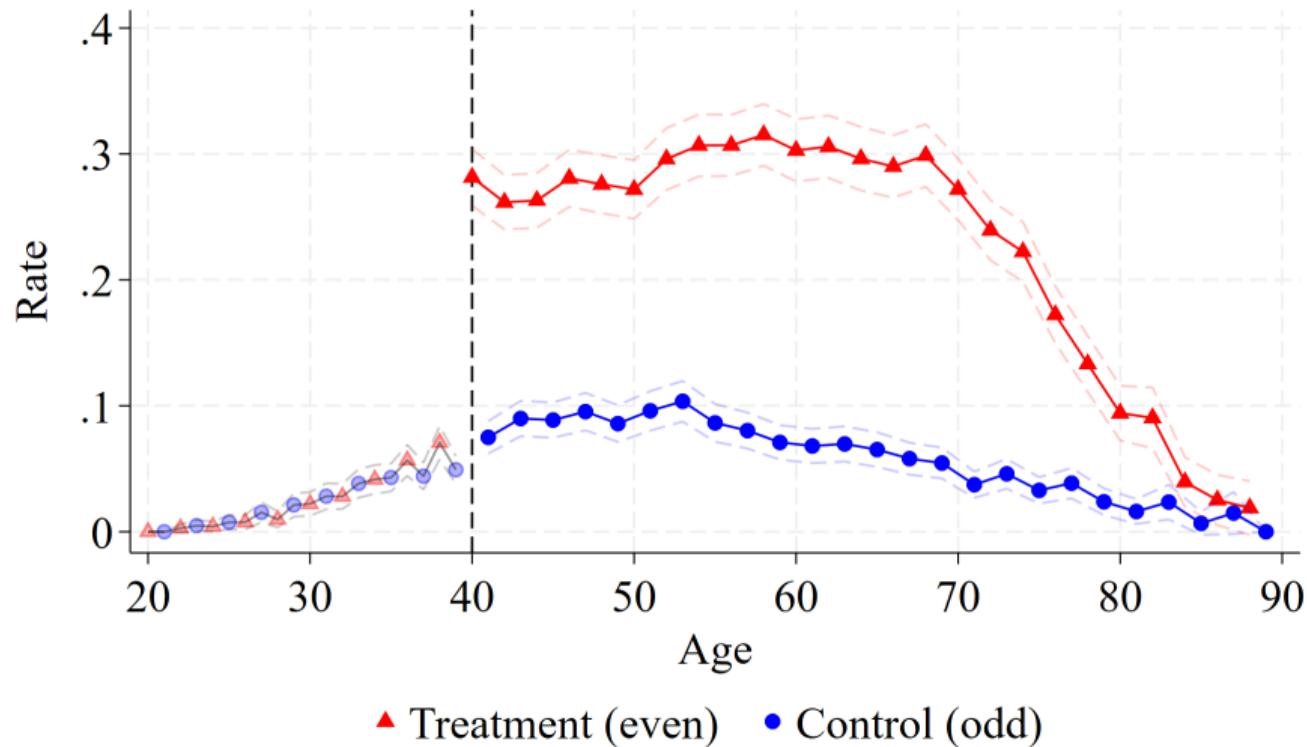
Effect on take-up

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# Breast screening



## Effect on 1-year take-up

	(1)	(2)	(3)	(4)
	General	Stomach	Breast	Cervical
treat	0.187*** (0.003)	0.190*** (0.003)	0.191*** (0.004)	0.164*** (0.003)
N	107,183	107,183	56,923	56,923
Control group mean	0.102	0.083	0.067	0.056
Percentage increase	183	229	283	295

## Effect on 1-year take-up

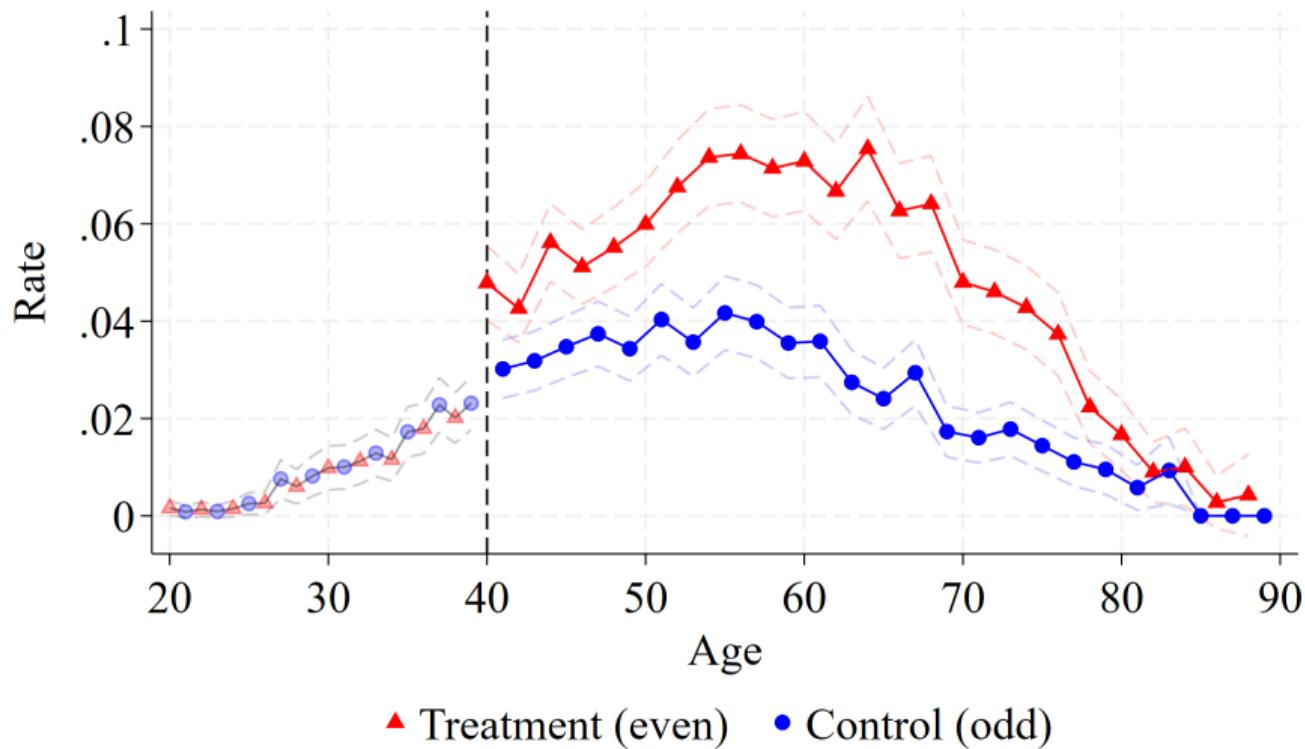
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- Arc elasticity: -0.48 to -0.72
- Arc elasticities from RAND HIE: -0.17 to -0.43 ([Newhouse, 1993](#))
- Arc elasticity from another study on screening subsidies: -0.47 ([Kim and Lee, 2017](#))
- Percentage increase in Oregon Medicaid experiment: 45-63% ([Finkelstein et al., 2012](#))

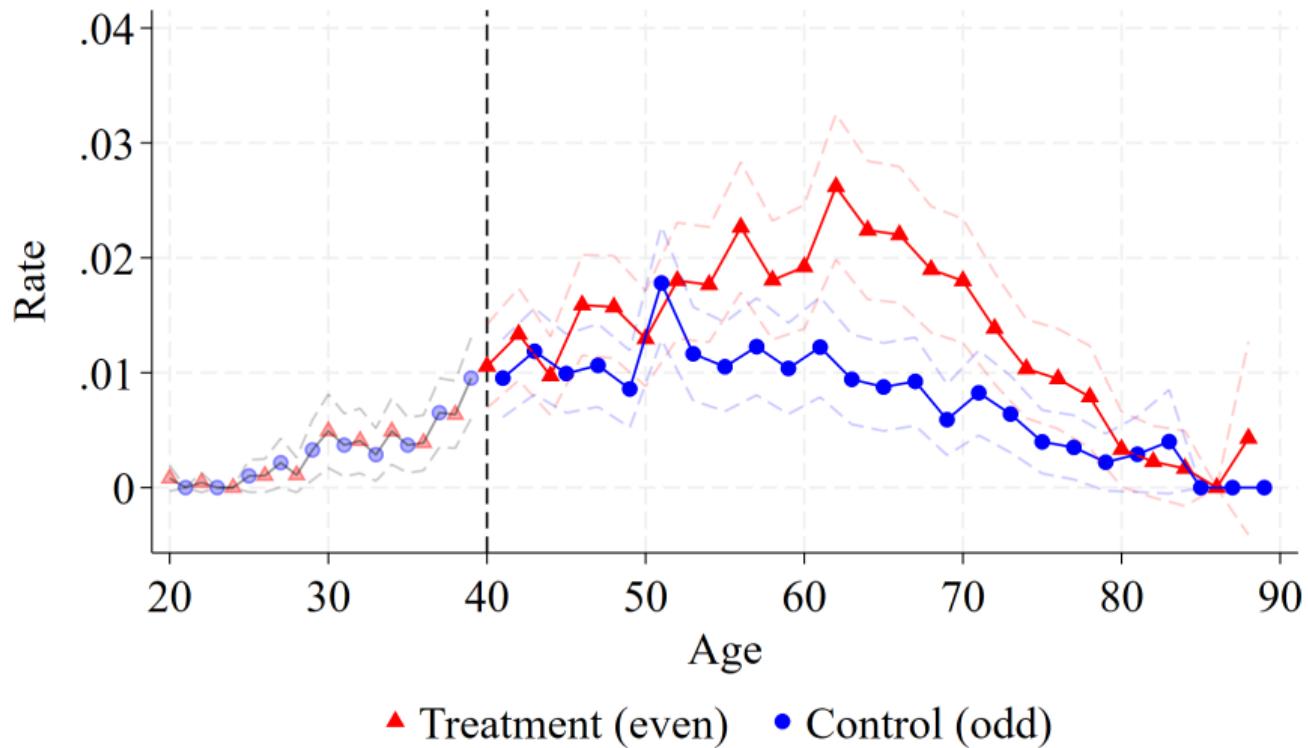
net increase

substitution

## Liver screening - subsidized every year



## Lung screening - not subsidized



## Cross-screening spillover

	(1)	(2)	(3)	(4)
	Annually subsidized		Unsubsidized	
	Liver	Colorectal	Prostate	Lung
treat	0.027*** (0.001)	0.033*** (0.001)	0.007*** (0.001)	0.0062*** (0.0007)
N	107,183	107,183	50,260	107,183
Control group mean	0.028	0.027	0.009	0.009
Percentage increase	94	124	81	67

# Cross-screening spillover

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- Receiving multiple screenings during a single hospital visit
- Cross-vaccine spillover: 8-34% ([Carpenter and Lawler, 2019](#)), 1.5% ([Humlum et al., 2024](#))

Same day share

Women

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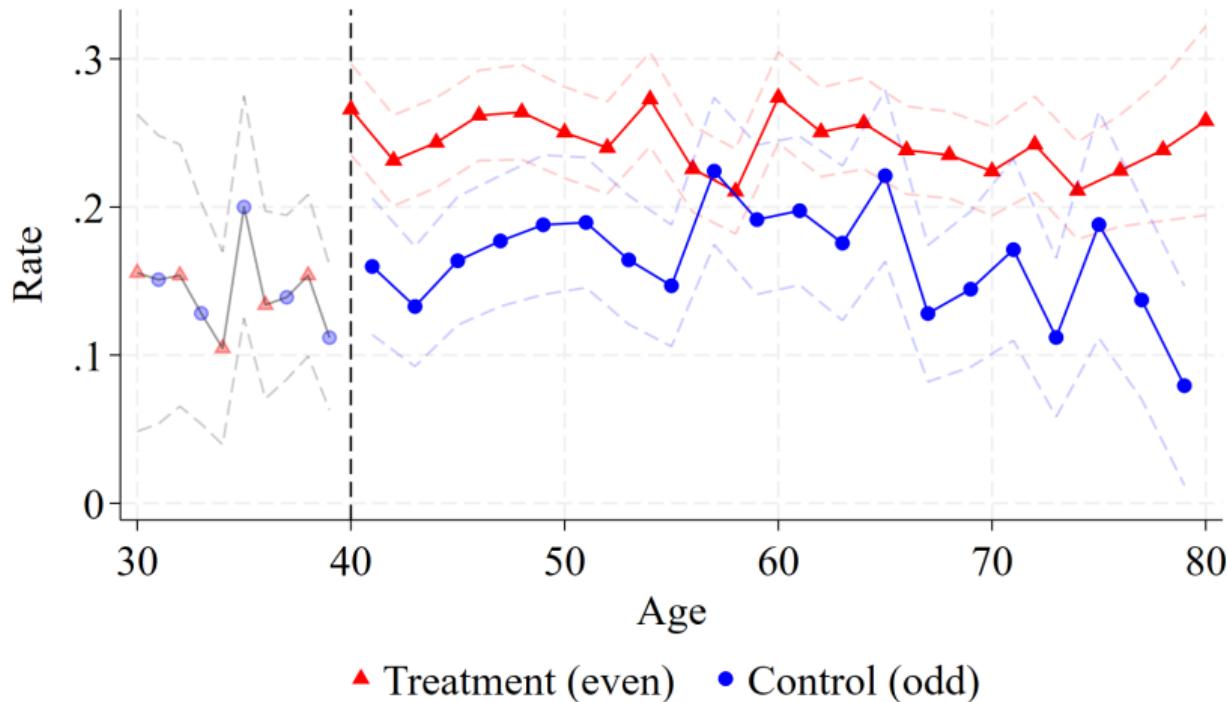
# Characterizing marginal participants (compliers)

- Who are marginal participants?

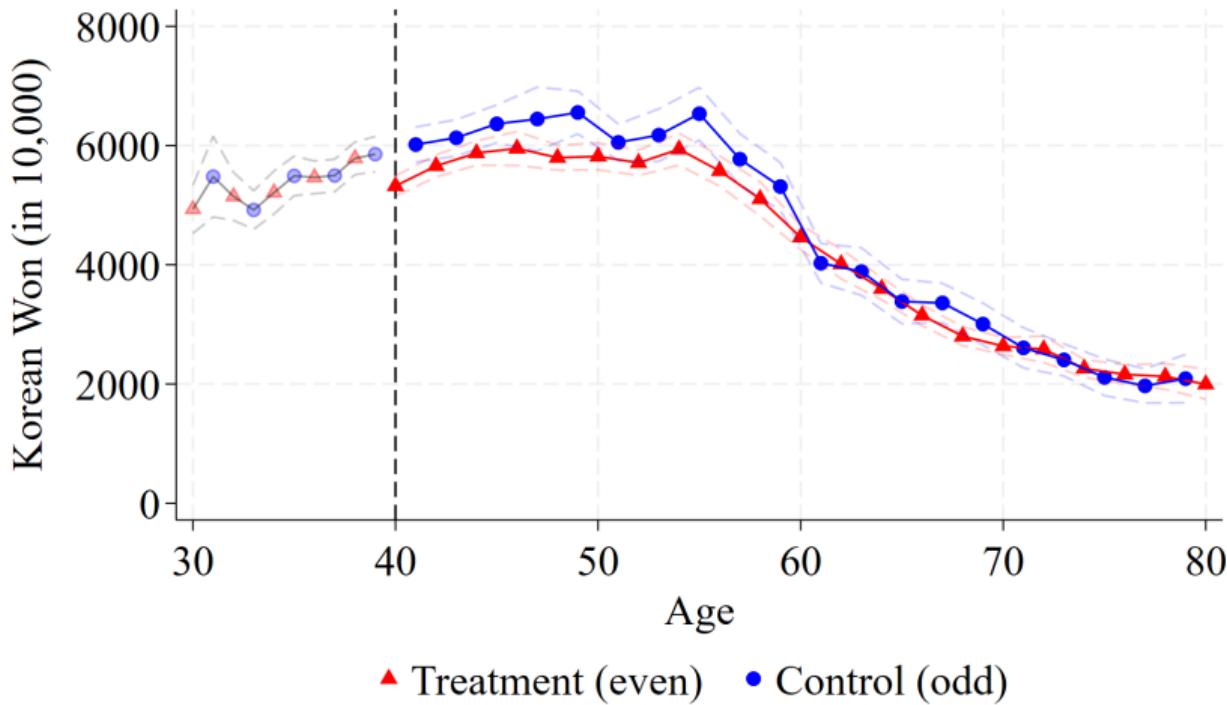
	Treatment (even)	Control (odd)
Always-takers	Yes	Yes
Compliers	Yes	No
Never-takers	No	No

- Cross-sectional & panel approach
- 3 steps for characterizing compliers
  - Participants at odd ages are always-takers
  - Participants at even ages are always-takers or compliers
  - Back out complier characteristics using group shares

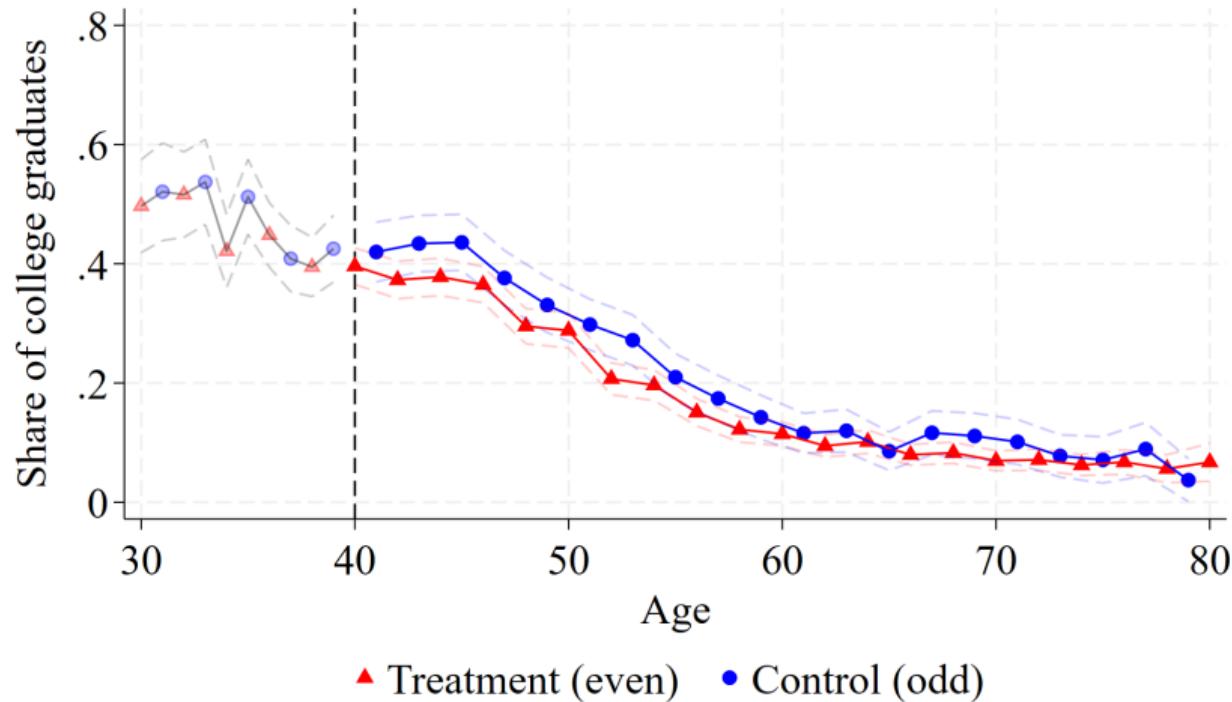
## Compliers are more likely to be diagnosed with a stomach disease



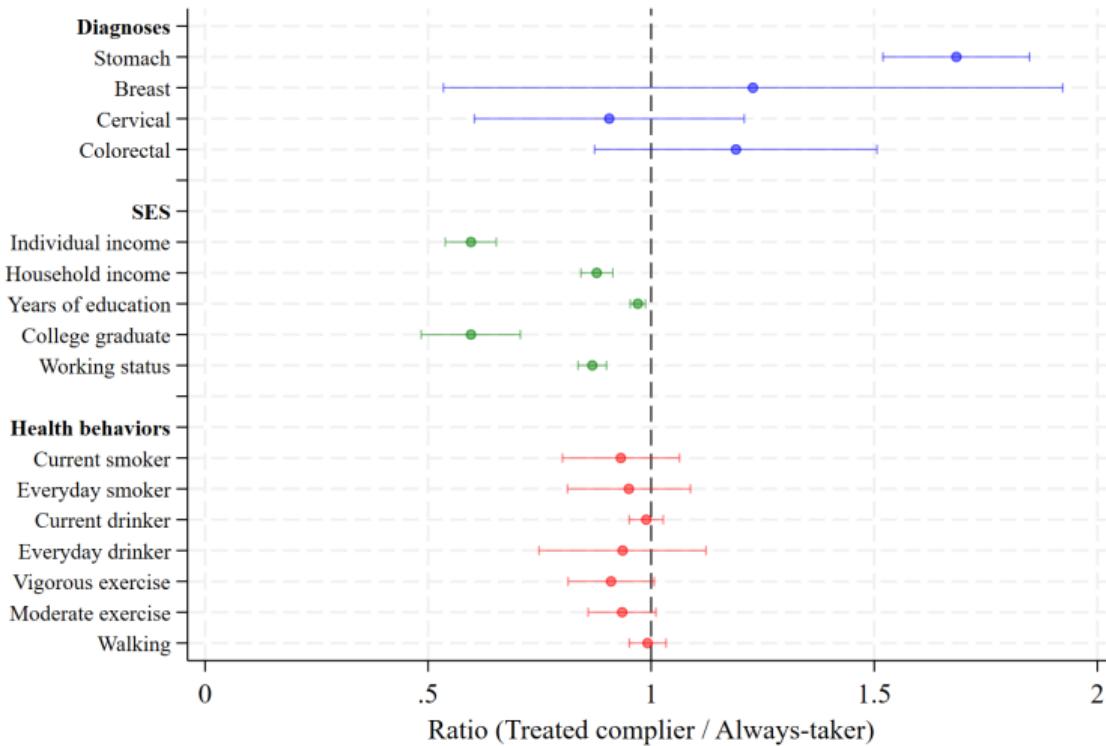
## Compliers have lower household income



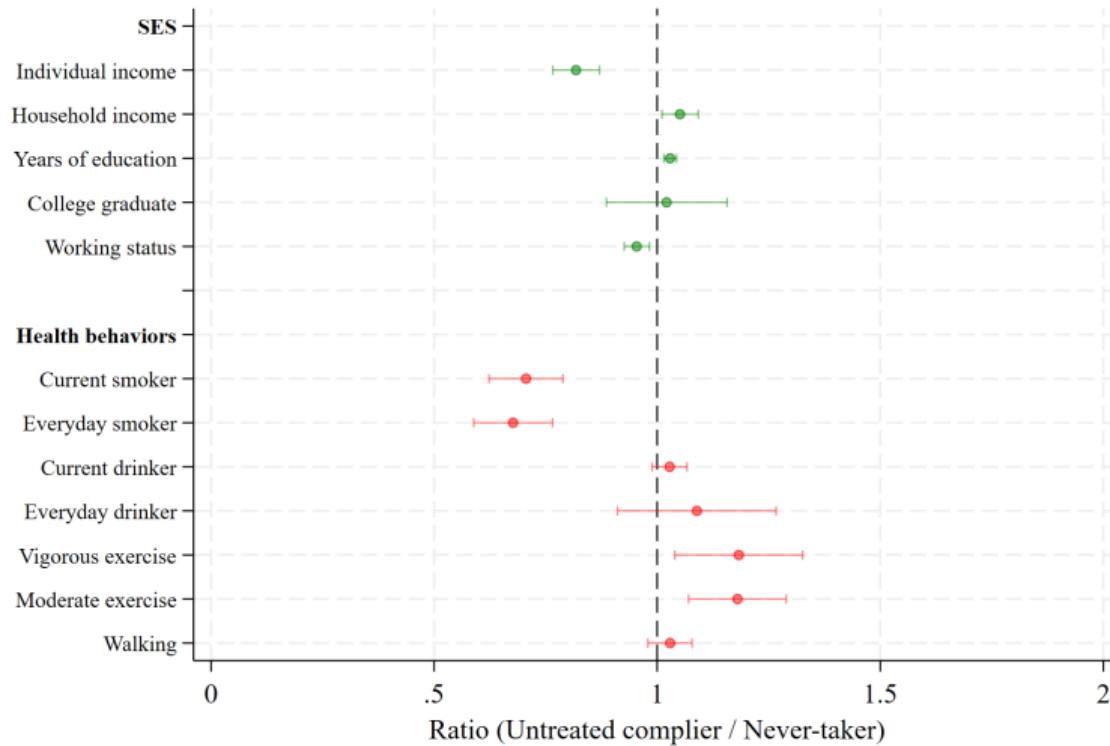
## Compliers are less likely to be college graduates



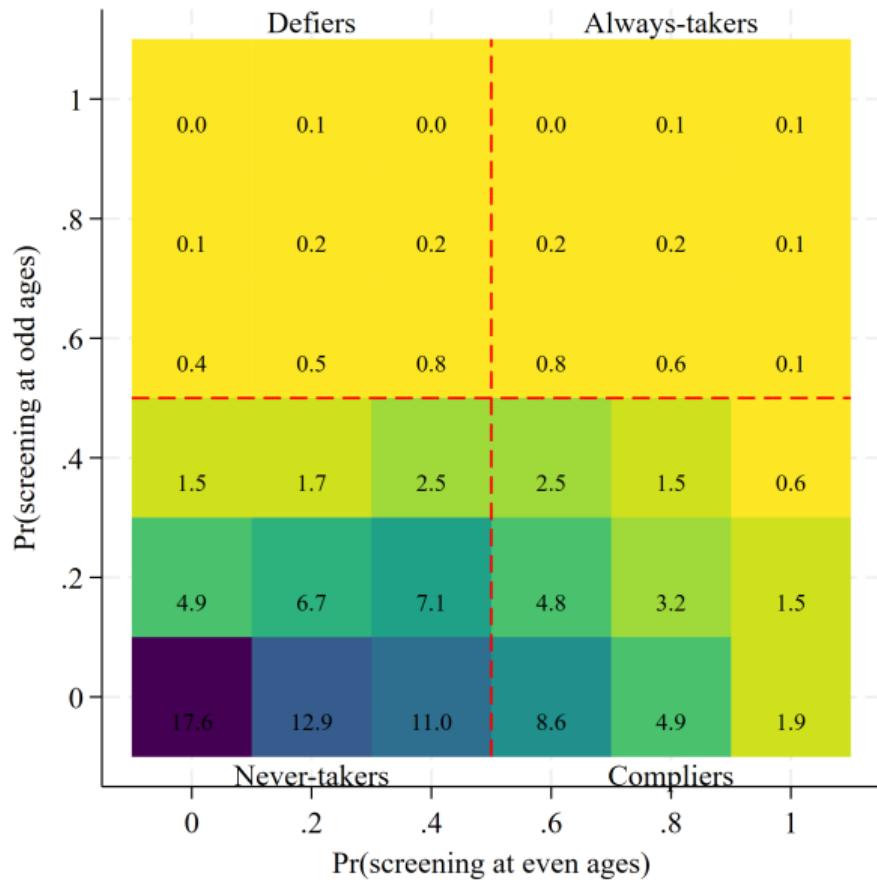
# Compliers vs Always-takers



# Compliers vs Never-takers



# Bivariate distribution of even and odd probabilities



# Compliers have poorer health

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Always	Compliers	Defiers	Never	Compliers / Always	Compliers / Defiers	Compliers / Never
<b>Panel A. Diagnoses</b>							
Stomach	0.181 (0.018)	0.256 (0.007)	0.189 (0.020)	-	1.413*** (0.143)	1.355** (0.146)	-
Breast	0.015 (0.007)	0.020 (0.003)	0.022 (0.011)	-	1.303 (0.653)	0.885 (0.452)	-
Cervical	0.061 (0.015)	0.058 (0.005)	0.058 (0.024)	-	0.950 (0.249)	0.997 (0.417)	-
Colorectal	0.183 (0.030)	0.260 (0.013)	0.212 (0.032)	-	1.423* (0.248)	1.226 (0.197)	-
Share	0.022	0.294	0.025	0.659			

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- Compliers through both **net increase channel** and through **intertemporal substitution channel** have poorer health

# Compliers are have lower SES

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Always	Compliers	Defiers	Never	Compliers / Always	Compliers / Defiers	Compliers / Never
<b>Panel B. SES</b>							
Individual income	2688 (247)	1043 (37)	2405 (222)	1288 (28)	0.388*** (0.038)	0.434*** (0.043)	0.810*** (0.034)
Household income	6093 (308)	3999 (67)	5587 (284)	3764 (46)	0.656*** (0.035)	0.716*** (0.038)	1.063*** (0.022)
Years of education	12.184 (0.312)	9.876 (0.098)	11.486 (0.349)	9.585 (0.075)	0.811*** (0.022)	0.860*** (0.027)	1.030** (0.013)
College graduate	0.270 (0.039)	0.123 (0.008)	0.291 (0.038)	0.142 (0.006)	0.456*** (0.072)	0.423*** (0.062)	0.869** (0.066)
Working status	0.753 (0.030)	0.560 (0.010)	0.733 (0.032)	0.603 (0.007)	0.744*** (0.033)	0.764*** (0.036)	0.930*** (0.020)

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# Compliers show better health behaviors than never-takers

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Always	Compliers	Defiers	Never	Compliers / Always	Compliers / Defiers	Compliers / Never
<b>Panel C. Health behaviors</b>							
Current smoker	0.272 (0.043)	0.268 (0.016)	0.381 (0.047)	0.399 (0.010)	0.984 (0.165)	0.702*** (0.096)	0.670*** (0.043)
Everyday smoker	0.260 (0.042)	0.257 (0.015)	0.350 (0.046)	0.386 (0.010)	0.991 (0.170)	0.735** (0.106)	0.667*** (0.044)
Current drinker	0.828 (0.042)	0.775 (0.014)	0.856 (0.031)	0.763 (0.008)	0.935 (0.050)	0.905*** (0.036)	1.016 (0.021)
Everyday drinker	0.111 (0.027)	0.141 (0.010)	0.175 (0.033)	0.151 (0.006)	1.274 (0.329)	0.807 (0.161)	0.933 (0.078)
Vigorous exercise	0.374 (0.032)	0.300 (0.010)	0.340 (0.028)	0.278 (0.006)	0.802*** (0.074)	0.883 (0.079)	1.080* (0.043)
Moderate exercise	0.526 (0.031)	0.474 (0.010)	0.525 (0.031)	0.419 (0.006)	0.902* (0.056)	0.903* (0.056)	1.133*** (0.029)
Walking	0.830 (0.023)	0.812 (0.007)	0.818 (0.019)	0.780 (0.005)	0.979 (0.029)	0.993 (0.025)	1.042*** (0.011)

# Summary of compliers' characteristics

- Compared to always-takers
  - Lower SES
  - Poorer health ([Kim and Lee, 2017](#))
- Compared to never-takers
  - Better health behaviors ([Oster, 2020; Einav et al., 2020; Kowalski, 2023](#))
- Targeting higher-risk participants
  - Guidelines alone cannot target high-risk (screening, vaccine)  
([Einav et al., 2020; Lawler, 2020](#))
  - Lower prices help target high-risk (screening, vaccine, drug adherence)  
([Choudhry et al., 2011; Baicker et al., 2015; Kim and Lee, 2017; Chandra et al., 2024; Humlum et al., 2024](#))

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# Data and econometric specification

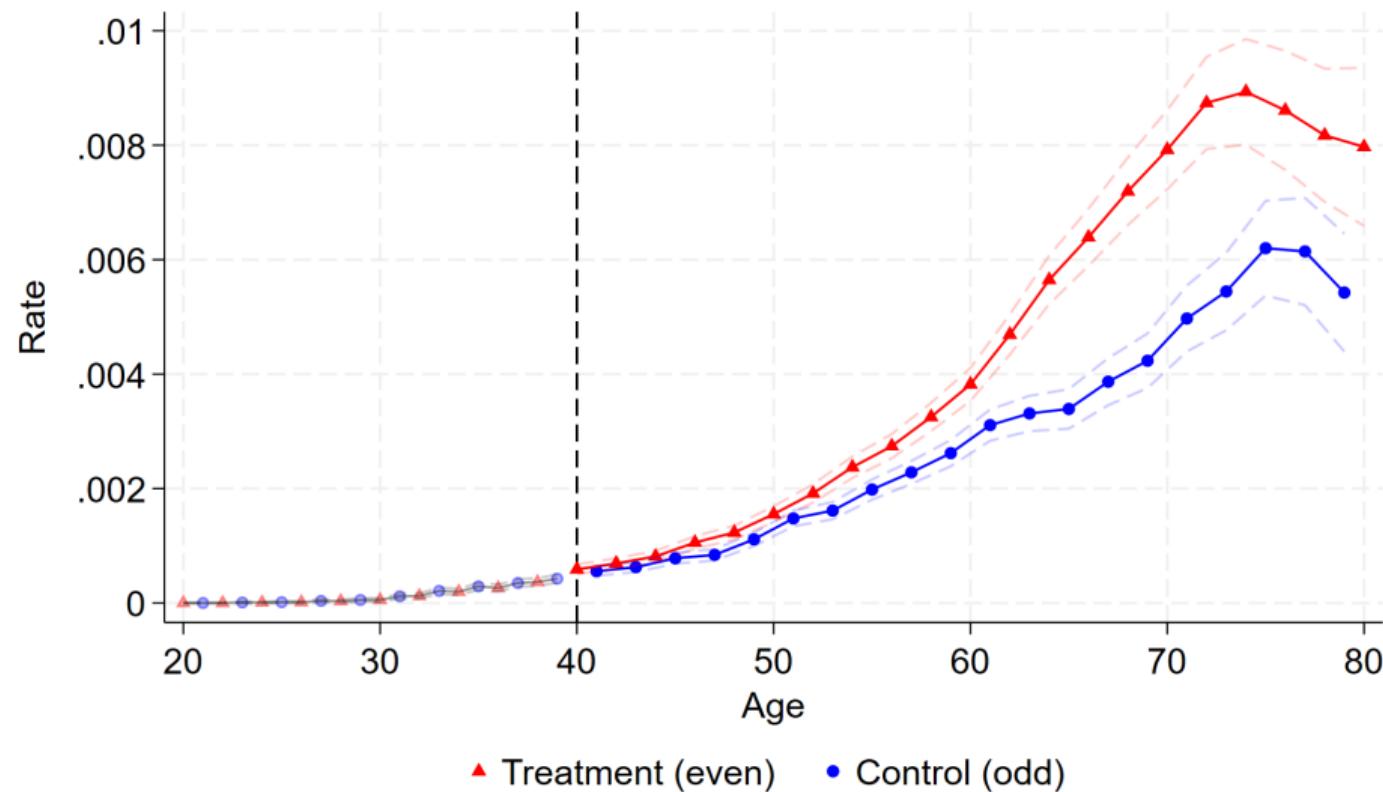
- NHIS claims data
  - All cancer diagnosis (true positive)
  - Preventive care use
  - Mortality

- Intent-to-treat (ITT) estimates

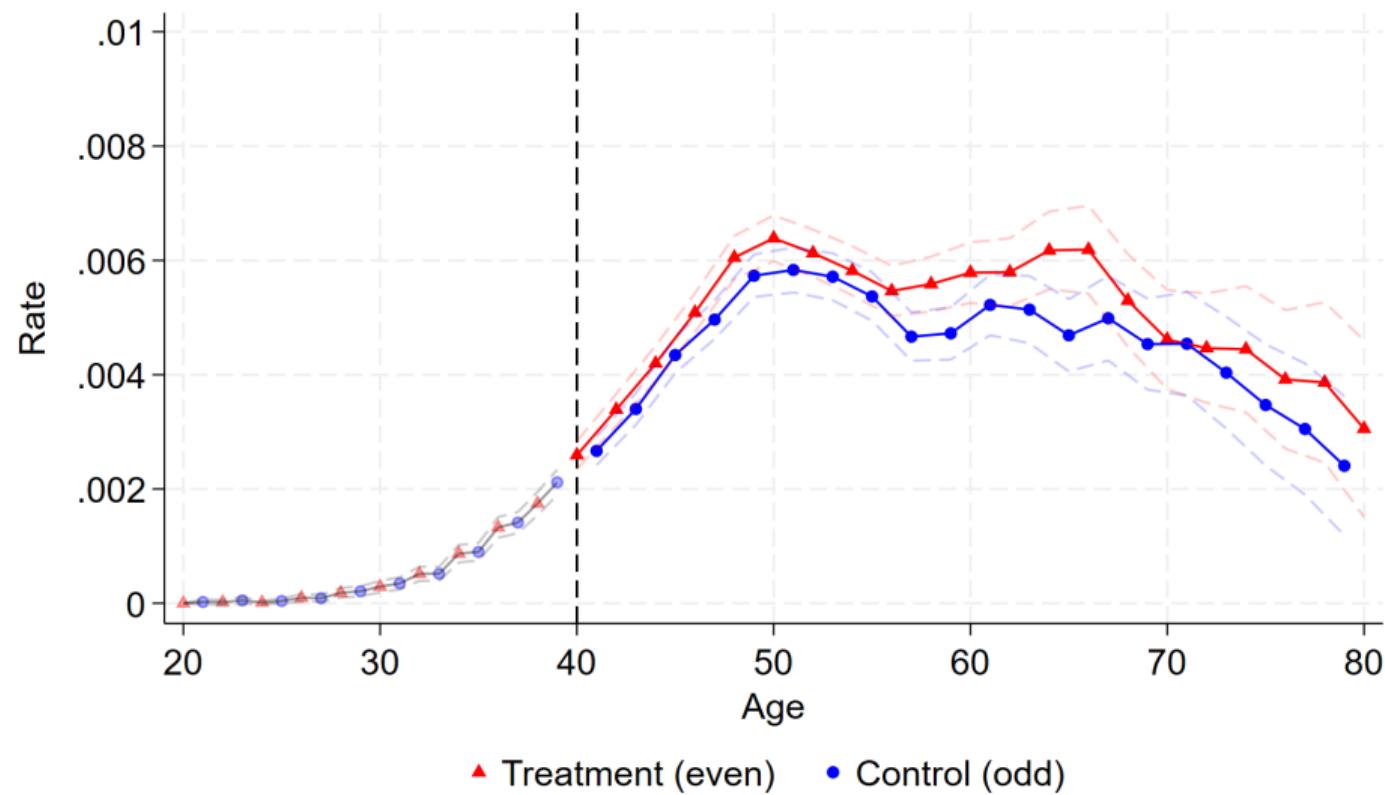
$$y_{it} = \eta_0 + \eta_1 \cdot treat_{it} + f(\text{age}_{it}) + \varepsilon_{it} \quad (2)$$

- Individual  $i$  in year  $t$  with age  $\in [40, 89]$
- $treat_{it} = 1$  if individual  $i$  turns even-numbered age in year  $t$
- $f(\text{age})$ : linear splines with 5 years interval
- Standard error clustered at the individual level
- Only capture short-run effect

## More stomach cancer diagnoses in the treatment group



## More breast cancer diagnoses in the treatment group



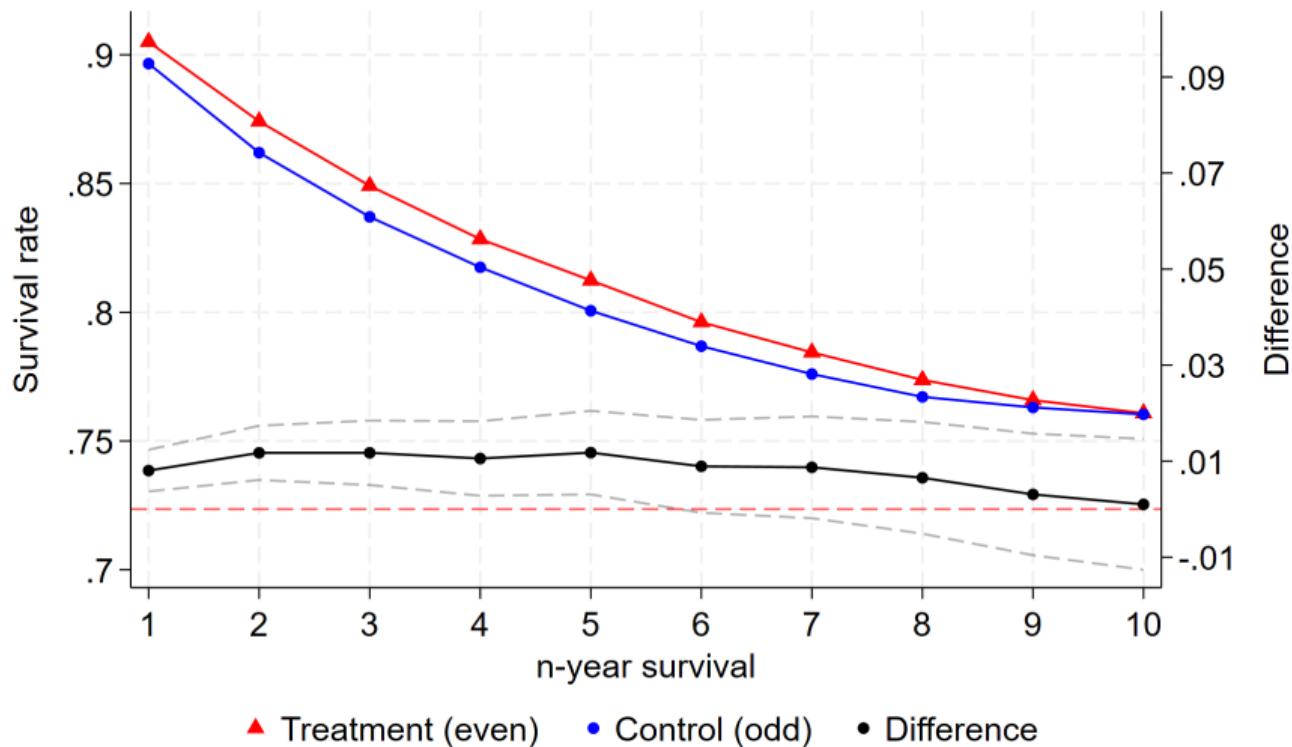
# Screening increases both **in-situ** and **invasive** cancer diagnoses

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
Any cancer	0.009	0.0016*** (0.0001)	18	7,449,256
invasive	0.008	0.0014*** (0.0001)	17	7,449,256
in-situ	0.001	0.0002*** (0.00003)	19	7,449,256
Stomach cancer	0.0020	0.00077*** (0.00006)	41	7,449,256
invasive	0.0018	0.00075*** (0.00005)	41	7,449,256
in-situ	0.0001	0.00003*** (0.000007)	55	7,449,256
Breast cancer	0.005	0.0005*** (0.0001)	11	3,503,656
invasive	0.004	0.0003*** (0.0001)	8	3,503,656
in-situ	0.001	0.00016*** (0.00004)	24	3,503,656
Cervical cancer	0.002	0.00020*** (0.00007)	13	3,503,656
invasive	0.001	0.00008 (0.00005)	11	3,503,656
in-situ	0.001	0.00014*** (0.00004)	17	3,503,656

## LATE comparison with other studies

- Local Average Treatment Effect = ITT (insurance claims) / First stage (survey)
  - Cancer detection rate among compliers
- LATE for breast screening: 0.0026
  - Organized breast screening program in Europe: 0.0010 (Guthmuller et al., 2023)
  - US mammogram recommendation at age 40: 0.0019 (Kadiyala and Strumpf, 2016)
  - Another mammogram subsidy in Korea: 0.0025 (Kim and Lee, 2017)
- LATE for stomach screening: 0.0041
  - Another stomach screening subsidy in Korea: 0.0025 (Kim and Lee, 2017)

# Cancer patients diagnosed in the treatment group live longer



## Screening leads to treatment of cancer precursors

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
Helicobacter Pylori	0.00017	0.00003*** (0.00001)	15	8,673,954
Polypectomy	0.01174	0.00724*** (0.00009)	62	8,673,954

# Screening increases medication use for chronic diseases

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
High blood pressure	0.23884	0.00046*** (0.00013)	0.2	8,673,954
Diabetes	0.08918	0.00015** (0.00006)	0.2	8,673,954
High cholesterol	0.13901	0.00419*** (0.00011)	3	8,673,954

High blood pressure   Diabetes   High cholesterol

## Screening increases medication use for osteoporosis

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
Tuberculosis	0.00195	0.00001 (0.00002)	0.5	8,673,954
Osteoporosis	0.02613	0.00106*** (0.00007)	4	8,673,954

# Impact on 1-year mortality

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
Total death	0.009084	-0.000048 (0.000072)	-0.530	6,740,415
Cancer death	0.002805	-0.000015 (0.000041)	-0.548	6,740,415
Stomach cancer death	0.000334	-0.000002 (0.000014)	-0.515	6,740,415
Breast cancer death	0.000153	-0.000011 (0.000013)	-7.057	3,480,294
Cervical cancer death	0.000059	0.000008 (0.000008)	12.735	3,480,294
Liver cancer death	0.000434	-0.000015 (0.000016)	-3.511	6,740,415
Colorectal cancer death	0.000287	0.000016 (0.000013)	5.715	6,740,415
Lung cancer death	0.000646	-0.000003 (0.000020)	-0.503	6,740,415
Prostate cancer death	0.000116	0.000003 (0.000012)	2.358	3,260,121

## Robustness checks

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## Takeaways

1. Subsidies can effectively target high-risk participants
2. Screenings significantly increase early detection of diseases and preventive care use

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## References

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# Implementation of the health screening program

- Nationwide program
  - Target: all the citizens covered by National Health Insurance Service (NHIS)
  - History: (1980) Beginning of the program → (2004) Current system
- How can I receive subsidized screenings
  - Providers: public health clinics / private clinics and hospitals designated by the NHIS  
⇒ (Dec 2023) 6,600 screening centers for general screening → 4600 people per center
  - Appointment: normally required but varies by hospitals and type of screenings
- Do people know about the screening program and the subsidies?
  - Even-odd subsidy rule has been used throughout the study period
  - Reminder mails (and mobile notifications)
    - Sent to those eligible for subsidies
    - Mail contains the type of screenings to receive and screening providers in the neighborhood

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# High risk group for liver screening

1. Individuals with the following diseases
  - Cirrhosis
  - Chronic liver disease
2. Individuals who were diagnosed with positive results in the previous year general health screening
  - Hepatitis B surface antigen test
  - Hepatitis C virus HCV antibody test

⇒ can be found through blood test
3. Individuals who used medical services for the following diseases in the past two years are excluded
  - Liver cell carcinoma, hepatocellular carcinoma and liver cancer (C22.0)
  - Intrahepatic bile duct carcinoma and Cholangiocarcinoma (C22.1)

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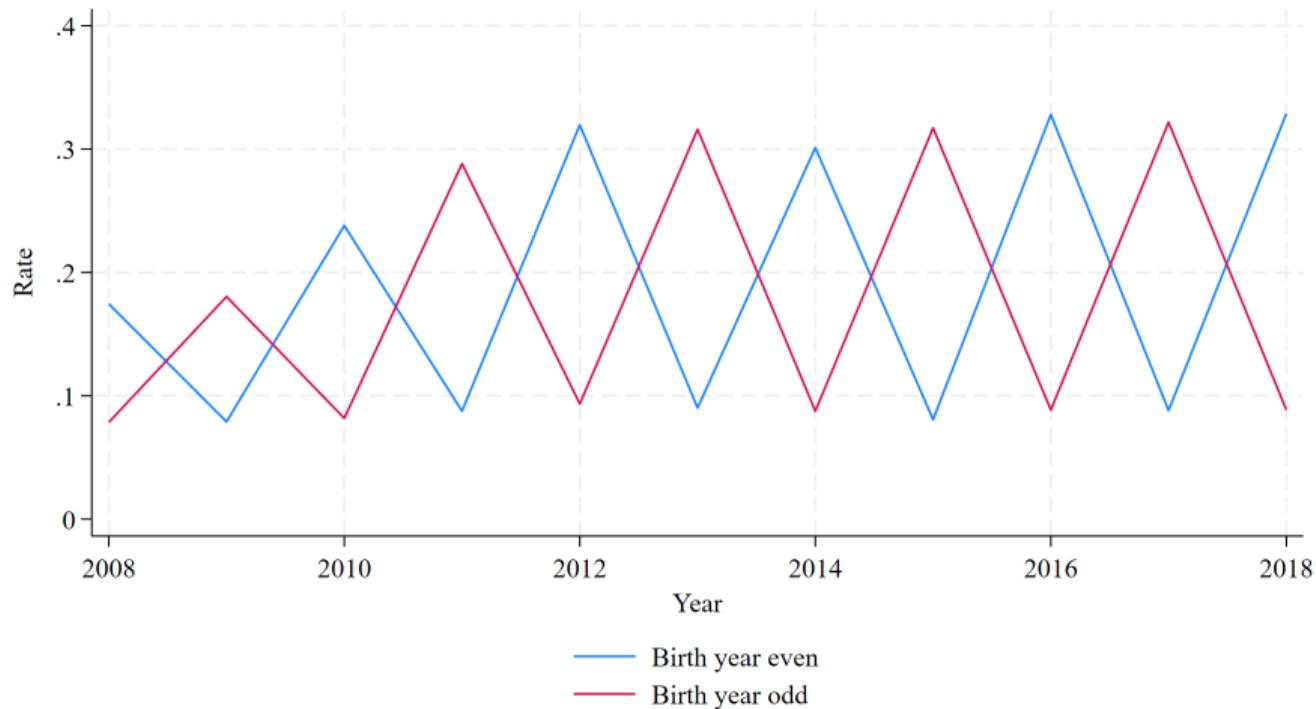
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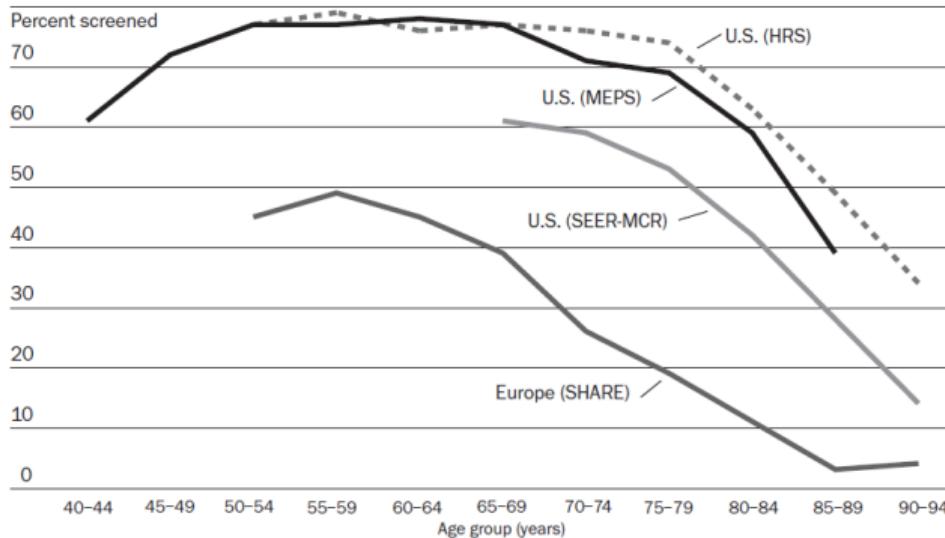
## Stomach screening by birth year



# Breast screening in the US and Europe (Howard et al., 2009)

## EXHIBIT 3

### Receipt Of Mammography In The Past Two Years Among Women Ages 44–94 In Europe And The United States, By Age Group, 2004

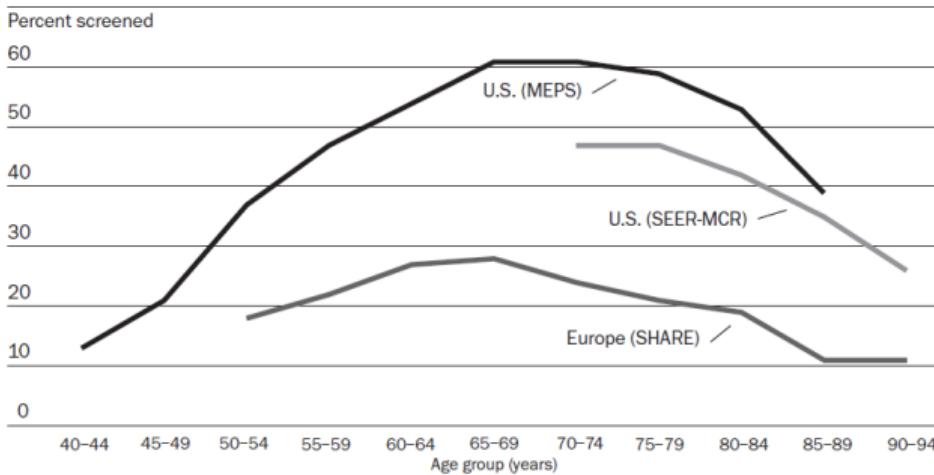


**SOURCES:** U.S. screening rates are from the Medical Expenditure Panel Survey (MEPS), the Health and Retirement Study (HRS), and Surveillance, Epidemiology, and End Results (SEER)-Medicare data (SEER-MCR). European rates are from the Survey of Health, Ageing, and Retirement in Europe (SHARE).

# Colorectal screening in the US and Europe (Howard et al., 2009)

## EXHIBIT 4

Receipt Of Colonoscopy, Sigmoidoscopy, And Fecal Occult Blood Tests Among Women And Men Ages 44-94 In The Past Ten Years In Europe And In The Past Five Years In The United States, By Age Group, 2004



**SOURCES:** U.S. screening rates are from the Medical Expenditure Panel Survey (MEPS); and Surveillance, Epidemiology, and End Results (SEER)-Medicare data (SEER-MCR). European rates are from the Survey of Health, Ageing, and Retirement in Europe (SHARE).

# Full balance table with unconditional difference

	(1)	(2)	(3)	(4)
	Even age group	Odd age group	Unconditional difference	Conditional difference
Age	58.697 (12.532)	59.240 (12.353)	-0.543*** (0.026)	- -
Female	0.530 (0.499)	0.532 (0.499)	-0.002** (0.001)	-0.002* (0.001)
Currently married	0.799 (0.401)	0.798 (0.402)	0.0009 (0.0009)	-0.0011 (0.0008)
Years of education	10.320 (4.510)	10.227 (4.538)	0.093*** (0.009)	-0.003 (0.008)
Working status	0.610 (0.488)	0.608 (0.488)	0.001 (0.002)	-0.003* (0.001)
Individual income	1446.3 (2081.6)	1425.7 (2068.1)	20.607*** (5.508)	2.762 (5.185)
Household income	4104.4 (3708.6)	4086.7 (3737.9)	17.735 (14.555)	3.221 (14.267)
Own a house	0.734 (0.442)	0.737 (0.441)	-0.002* (0.001)	-0.0002 (0.0011)
Number of household members	3.067 (1.317)	3.051 (1.317)	0.016*** (0.003)	-0.004 (0.003)
N	54274	52909		
Share	(0.51)	(0.49)		
F(8, 15939)				1.65 (0.10)

## Robustness check for balance table with different intervals

	(1)	(2)	(3)
	3 years	5 years	7 years
Female	-0.002*	-0.002*	-0.002*
	(0.001)	(0.001)	(0.001)
Currently married	-0.001*	-0.001	-0.001
	(0.001)	(0.001)	(0.001)
Years of education	-0.003	-0.003	-0.003
	(0.008)	(0.008)	(0.008)
Working status	-0.003**	-0.003*	-0.003*
	(0.002)	(0.001)	(0.001)
Individual income	1.1	2.8	1.2
	(5.3)	(5.2)	(5.2)
Household income	0.6	3.2	-4.5
	(15.4)	(14.3)	(14.1)
Own a house	-0.000	-0.000	0.000
	(0.001)	(0.001)	(0.001)
Number of household members	-0.004	-0.004	-0.004*
	(0.003)	(0.003)	(0.003)

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# Bounding estimates for balance check

	(1) Age ∈ [39, 89]	(2) Age ∈ [40, 89]	(3) Age ∈ [41, 89]
Age	0.521*** (0.026)	-0.543*** (0.026)	0.562*** (0.025)
Female	-0.001 (0.001)	-0.002** (0.001)	-0.001 (0.001)
Currently married	-0.0009 (0.0009)	0.0009 (0.0009)	-0.0018** (0.0009)
Years of education	-0.094*** (0.009)	0.093*** (0.009)	-0.107*** (0.010)
Working status	-0.006*** (0.001)	0.001 (0.002)	-0.007*** (0.002)
Individual income	-16.235*** (5.470)	20.607*** (5.508)	-24.789*** (5.618)
Household income	-23.153 (14.766)	17.735 (14.555)	-31.182** (14.995)
Own a house	0.003*** (0.001)	-0.002* (0.001)	0.003** (0.001)
Number of household members	-0.027*** (0.003)	0.016*** (0.003)	-0.034*** (0.003)
N	110121	107183	104153

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## Health care usage data collection

- Recording health care usage
  - Survey participants are asked to keep health diary and store receipts from every visit to hospitals and pharmacies
- No gap
  - During annual interviews, enumerator goes through health diary from the last time of interview

# Health diary

## ❶ 건강기계부 작성방법 ❶

### ❷ 병의원에 다녀왔을 때

- 우리 가족 누구든지 병의원에 다녀오면 가계부를 작성해주세요.
- 병원 영수증과 처방전 및 약국 영수증은 영수증 보관함에 함께 모아주세요.

(작성 예시) 이를 흥길동이 아버지에게 비염 때문에 다녀온 후

의료 이용 형태	<input checked="" type="checkbox"/> 외래	<input type="checkbox"/> 입원	<input type="checkbox"/> 응급	<input type="checkbox"/> 건강검진
진료 일	2019년 4월 10일(부터) ~ 11일(까지)			
가구원 이동	중길동			
병원 이름	평촌한 아버지병원			
방문 이유	알레르기 비염			
병원 수납금액	4,000 원			
교통수단 택시, 버스, 도보 등	내원	걸어서	자가	걸어서
보관여부	<input checked="" type="checkbox"/> 전화비 납입 영수증	<input type="checkbox"/> 처방전	<input checked="" type="checkbox"/> 약국영수증	

### ❸ 의약품 및 보건의료용품을 샀을 때

- 우리 가족 누구든지 처방전 없이 의약품 또는 의료기기, 건강기능식품 등을 구매하면 가계부에 기입해주세요.
- 다음과 같은 항목을 구매한 경우 월별로 합산하여 기입해주세요.  
☞ 구입영수증은 영수증 보관함에 따로 모아주세요.

(예시) 엄마와 함께 먹으려고 알약과 구입, 잡기 기분이 있어 엄마가 종합기장을 약국에서 구매

구입처	구입 장소	비용
1. 일반약/의약품 (제약/약국)	<input type="checkbox"/> 병의원 <input checked="" type="checkbox"/> 약국 <input type="checkbox"/> 마약환장	( 6,000 ) 원
2. 한약 및 한약대 (제약/한약재)	<input type="checkbox"/> 약국 <input type="checkbox"/> 한약방	( ) 원
3. 건강보조식품 (제약, 비타민 등)	<input type="checkbox"/> 병의원 및 약국 <input checked="" type="checkbox"/> 인터넷 및 품소형 <input type="checkbox"/> 백화점, 마트, 시장 등	( 47,500 ) 원
4. 의료기기 및 의료용품 ※ 예시		( ) 원
+ 보건의료소모품(밴드, 마스크, 속눈수, 연고, 모기제제 등) + 안경 및 관리제품(구입 및 수리) + 보청기 구입 및 수리 + 신체진찰 의료기기 등 기타 의료용품 구매, 대여 및 수리 (예제로, 안경 편광경, 막苦难, 속눈수 교체, 헬프기, 불량측정기 등)		

## <How to write health diary>

- Visit to hospital
  - Record it for all the household members
  - Store hospital receipts, prescriptions and pharmacy receipts in a box

<Example> After a visit to ENT for allergy

Type	<input checked="" type="checkbox"/> Outpatient	<input type="checkbox"/> Inpatient	<input type="checkbox"/> Emergency	<input type="checkbox"/> Screening
Date	From: April 10, 2019 To:			
Name	John Doe			
Name of the hospital	Dr. Jane M. Doe, MD			
Purpose	Allergy			
Hospital bills	\$40			
Transportation	To	Walking	From	Walking
Receipts	<input type="checkbox"/> Hospital	<input type="checkbox"/> Prescription	<input type="checkbox"/> Pharmacy	

- Purchase of OTC drugs, oriental medicine, dietary supplements
  - Record it for all the household members
  - Store hospital receipts, prescriptions and pharmacy receipts in a box

<Example> Purchase of multivitamin and Tylenol

January 2019		
Item	Place	Cost
OTC drugs	<input type="checkbox"/> Hospital <input type="checkbox"/> Pharmacy <input type="checkbox"/> CVS	( ) KRW ( ) KRW ( ) KRW
Oriental medicine	<input type="checkbox"/> Pharmacy <input type="checkbox"/> Acupuncture clinic	( ) KRW ( ) KRW
Dietary supplement (ginseng, vitamin, etc)	<input type="checkbox"/> Hospital or pharmacy <input type="checkbox"/> Internet shopping <input type="checkbox"/> Department store	( ) KRW ( ) KRW ( ) KRW
Any other medical products (e.g.)		( ) KRW
- Bandage, mask, insect repellent		
- Glasses, contact lenses		
- Hearing aid		

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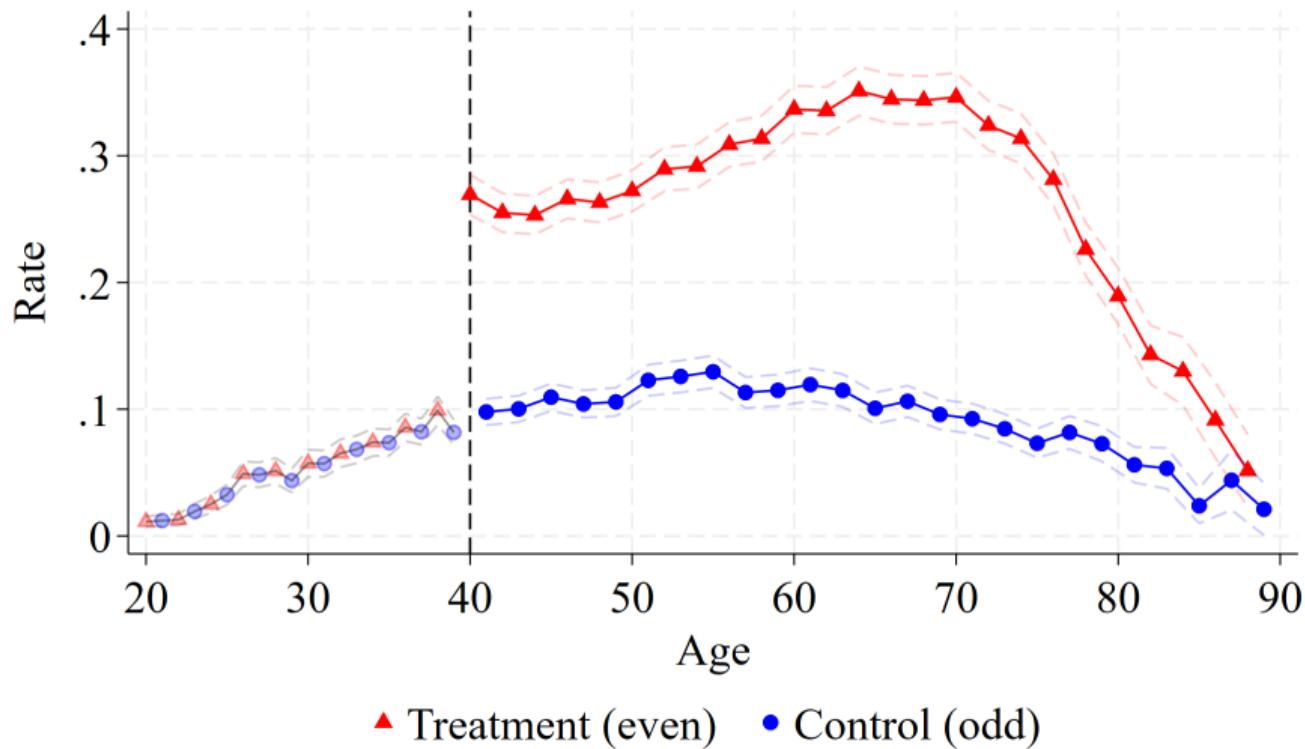
**Effect on take-up**

Cross spillover

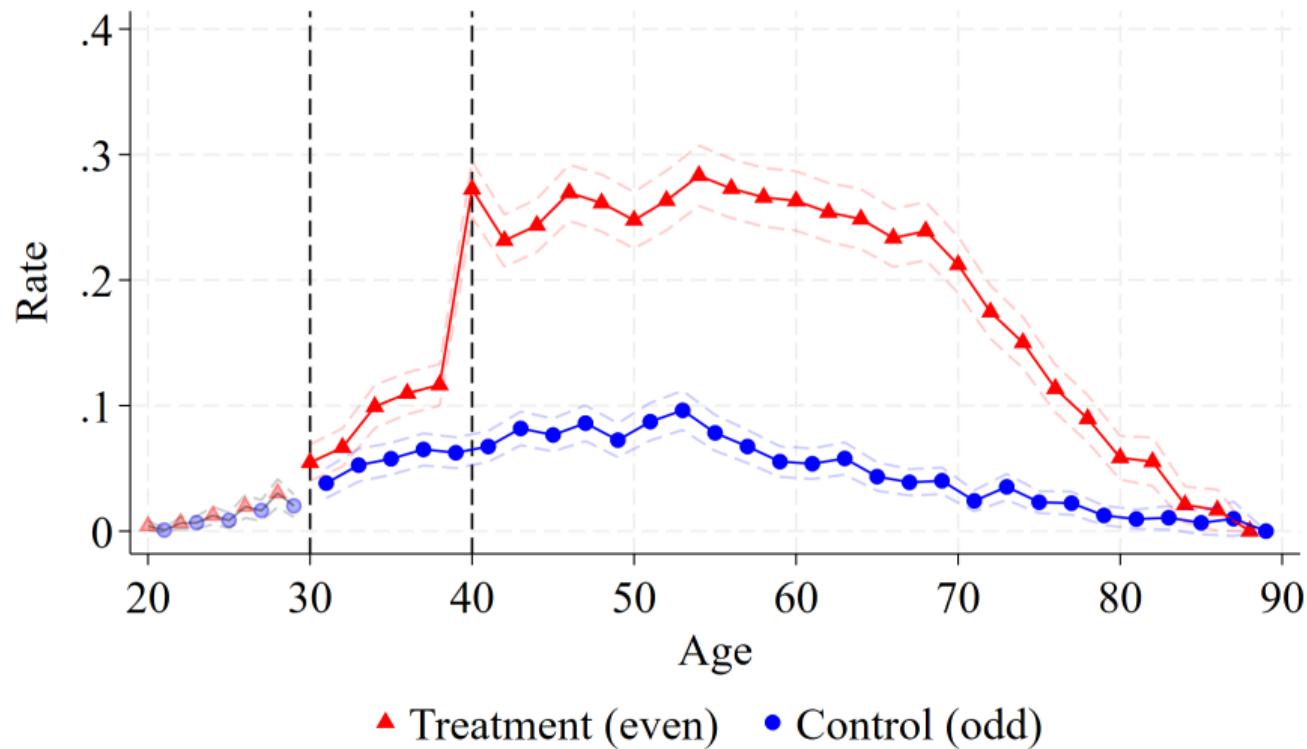
Selection

Effect

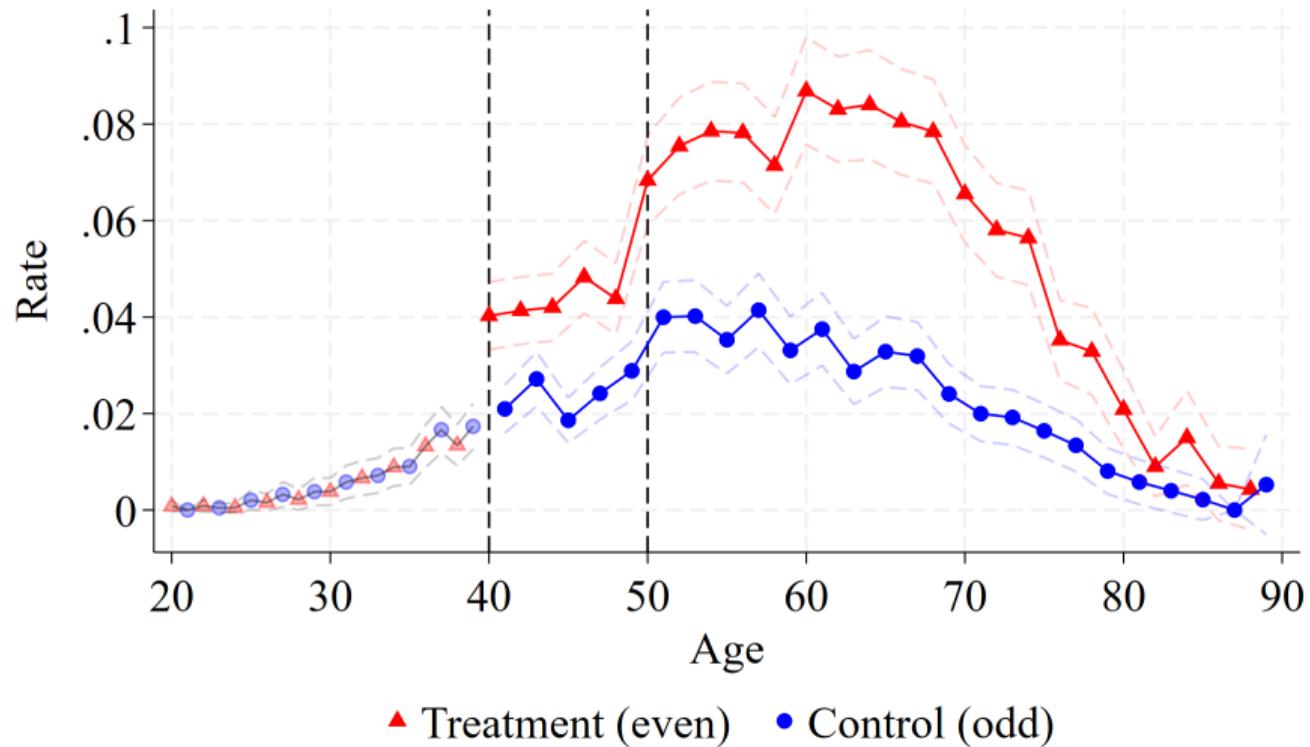
## General screening



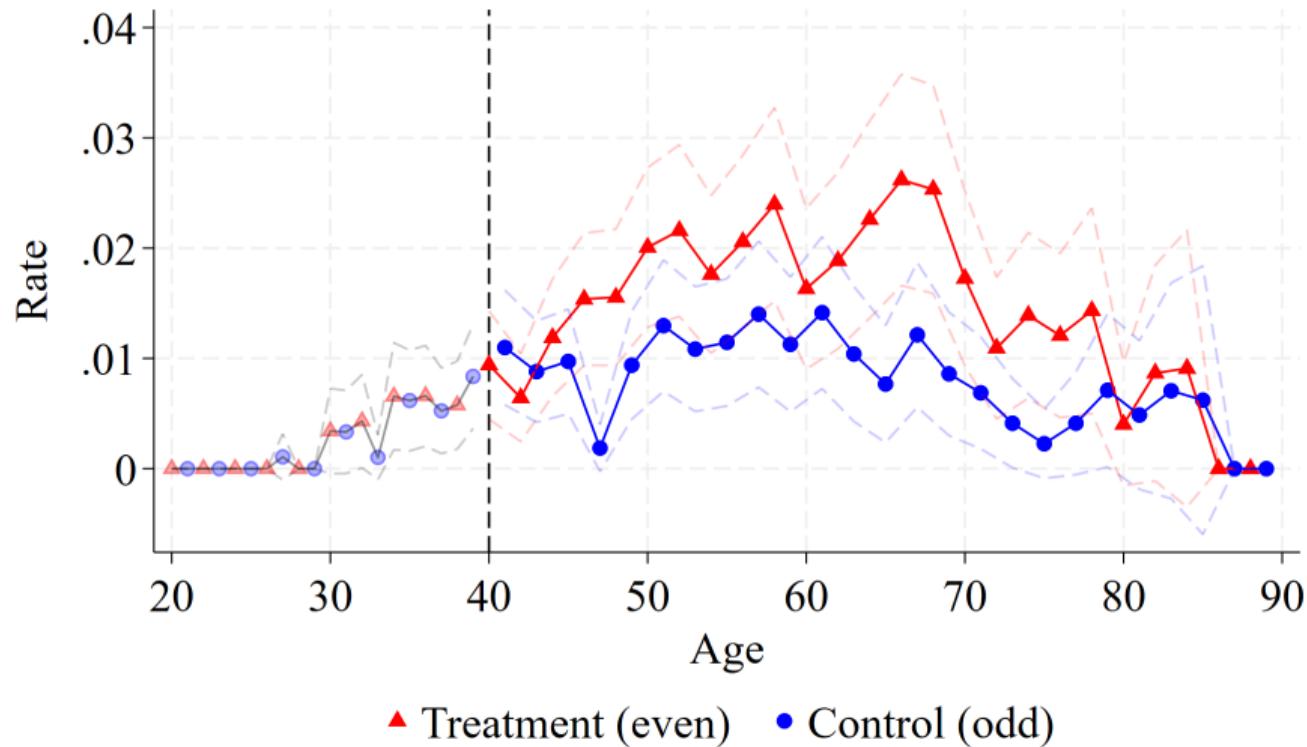
# Cervical screening



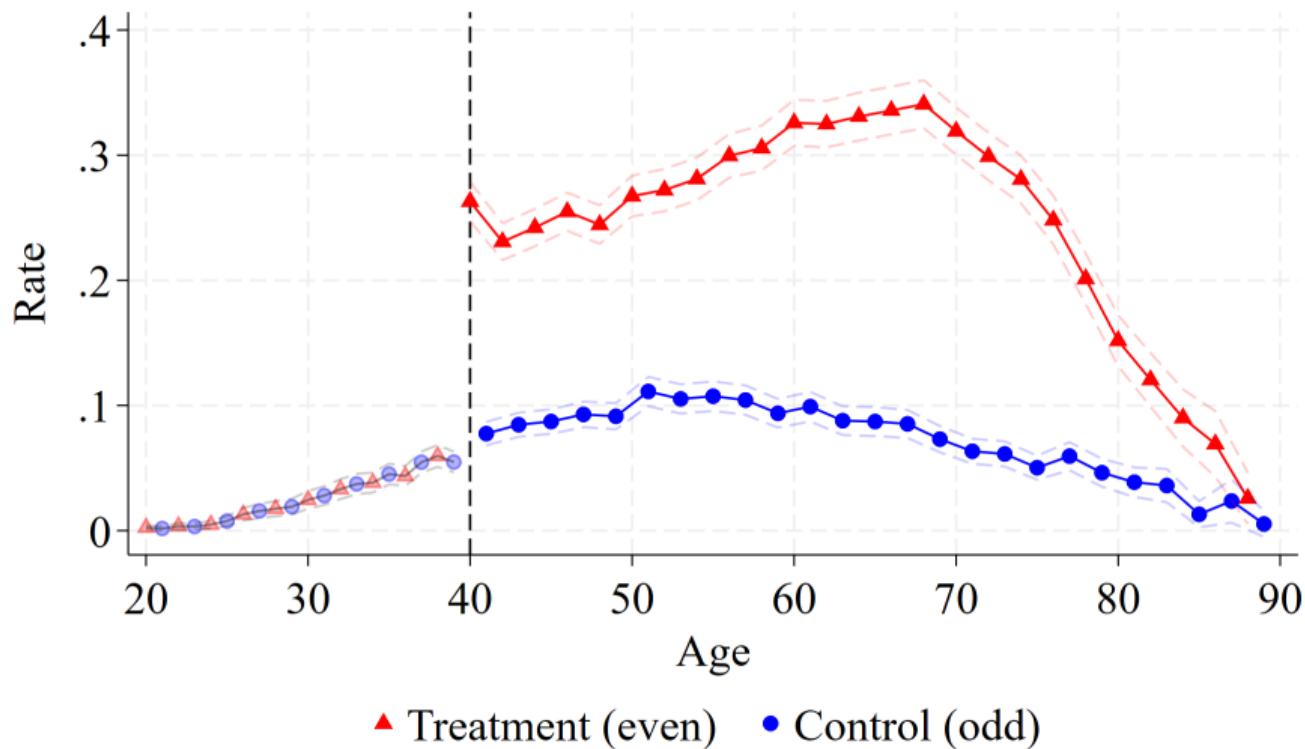
## Colorectal screening



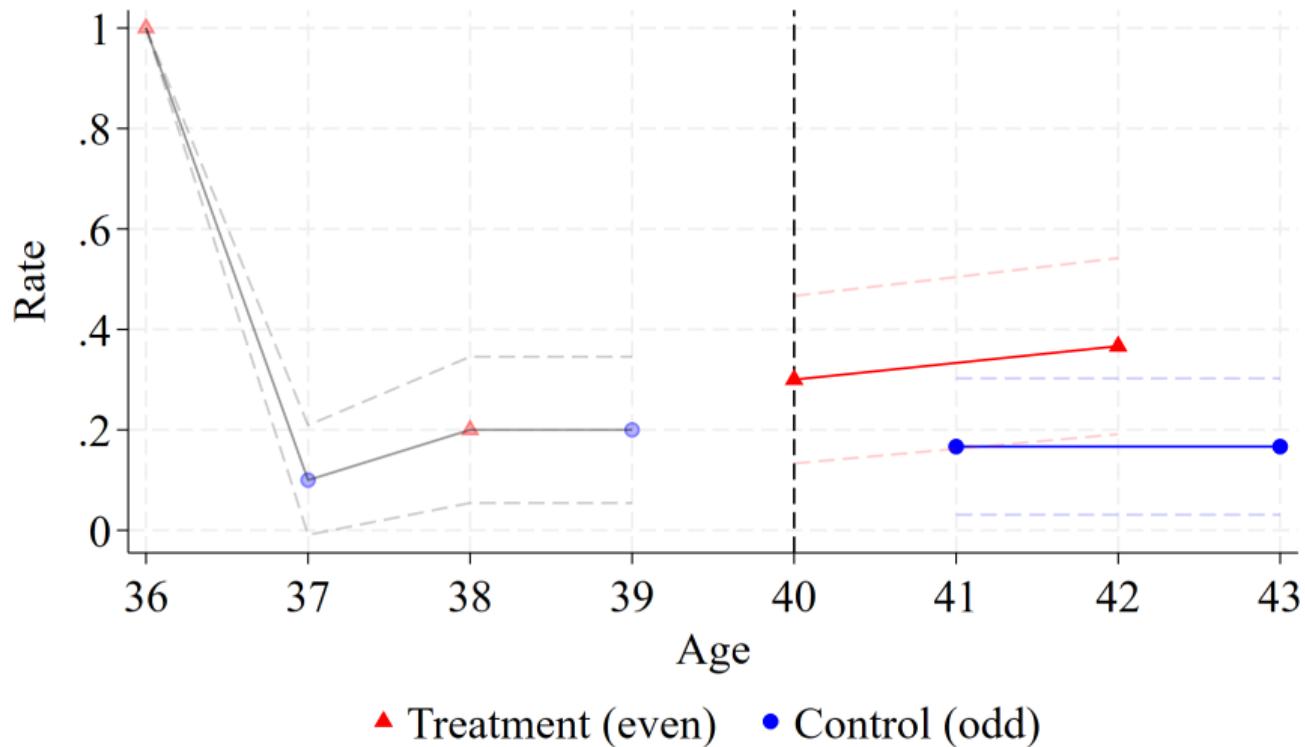
# Prostate screening



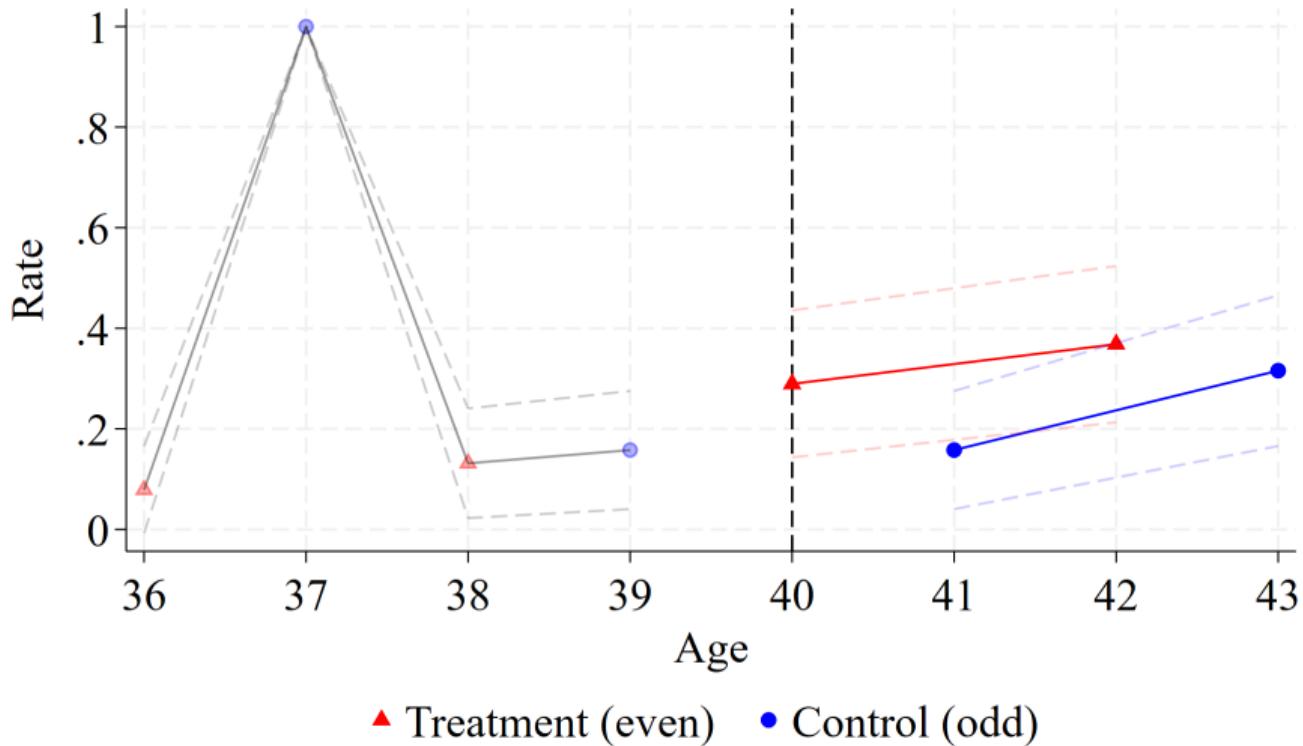
## Net increase in screening participation at age 40



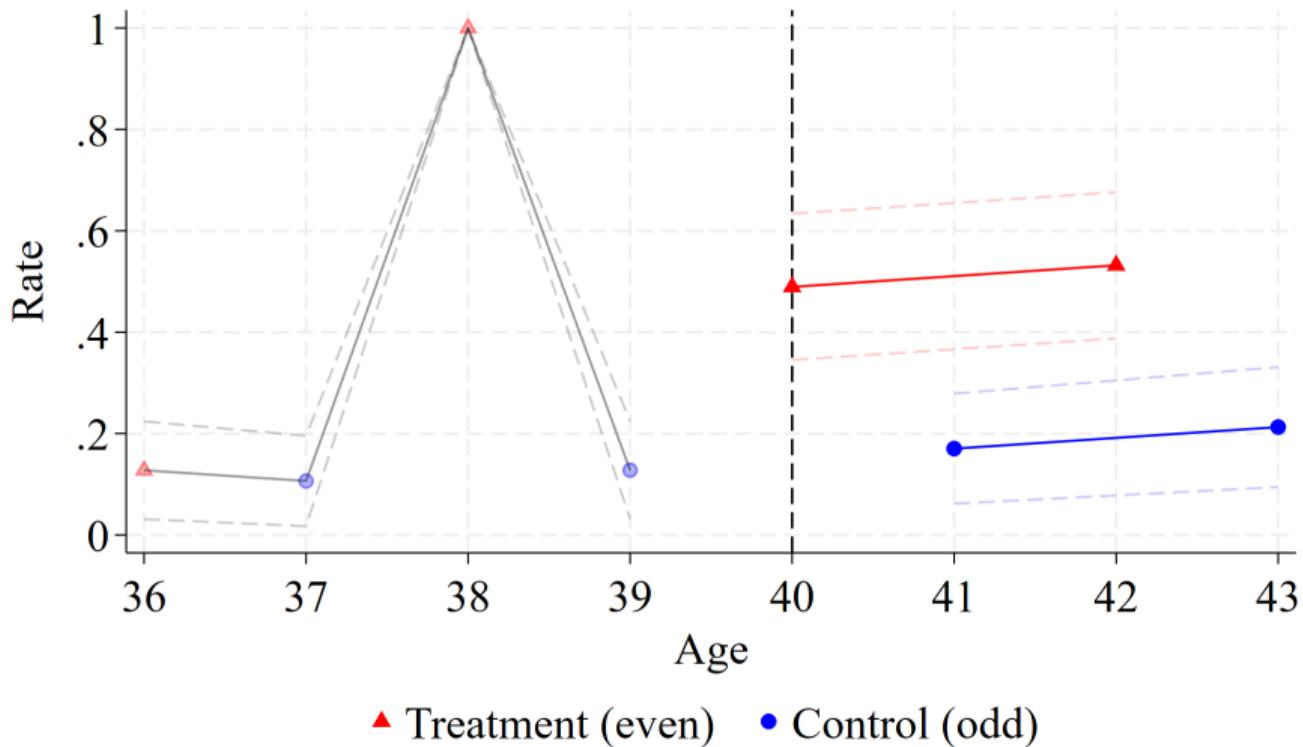
## No substitution for participants at age 36



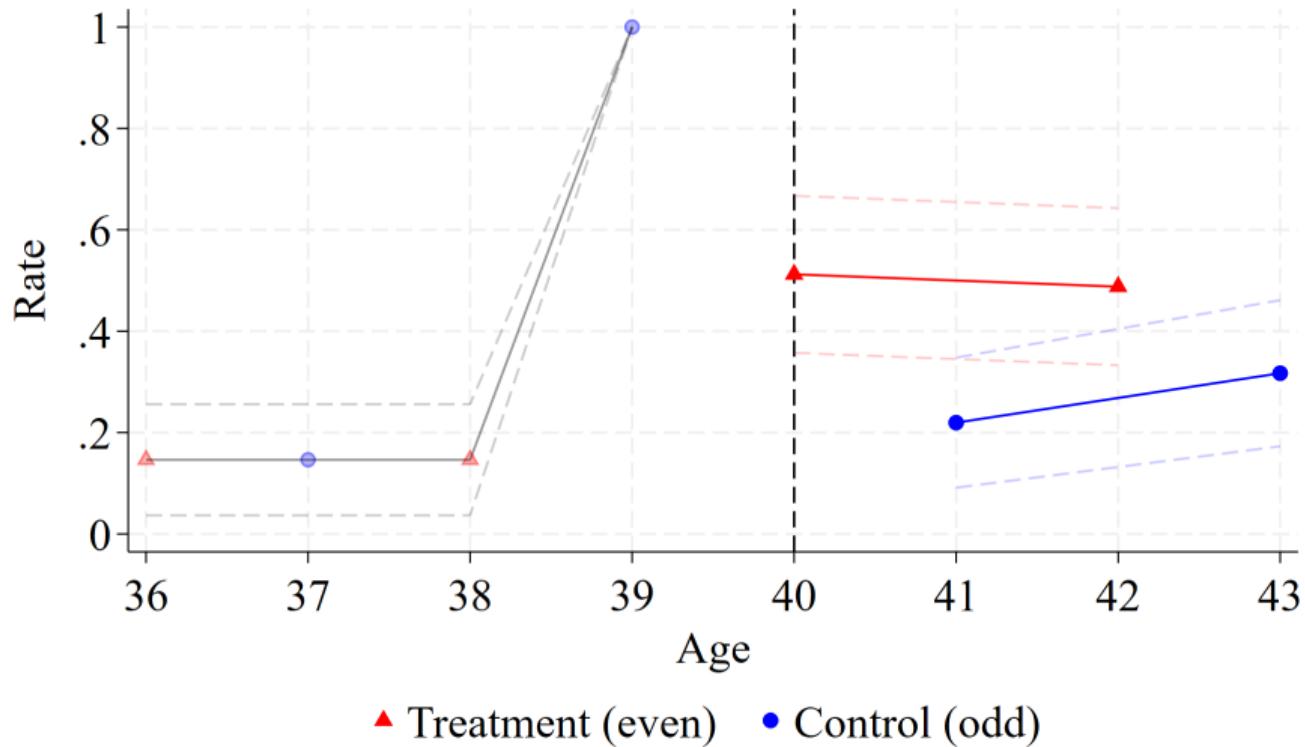
## No substitution for participants at age 37



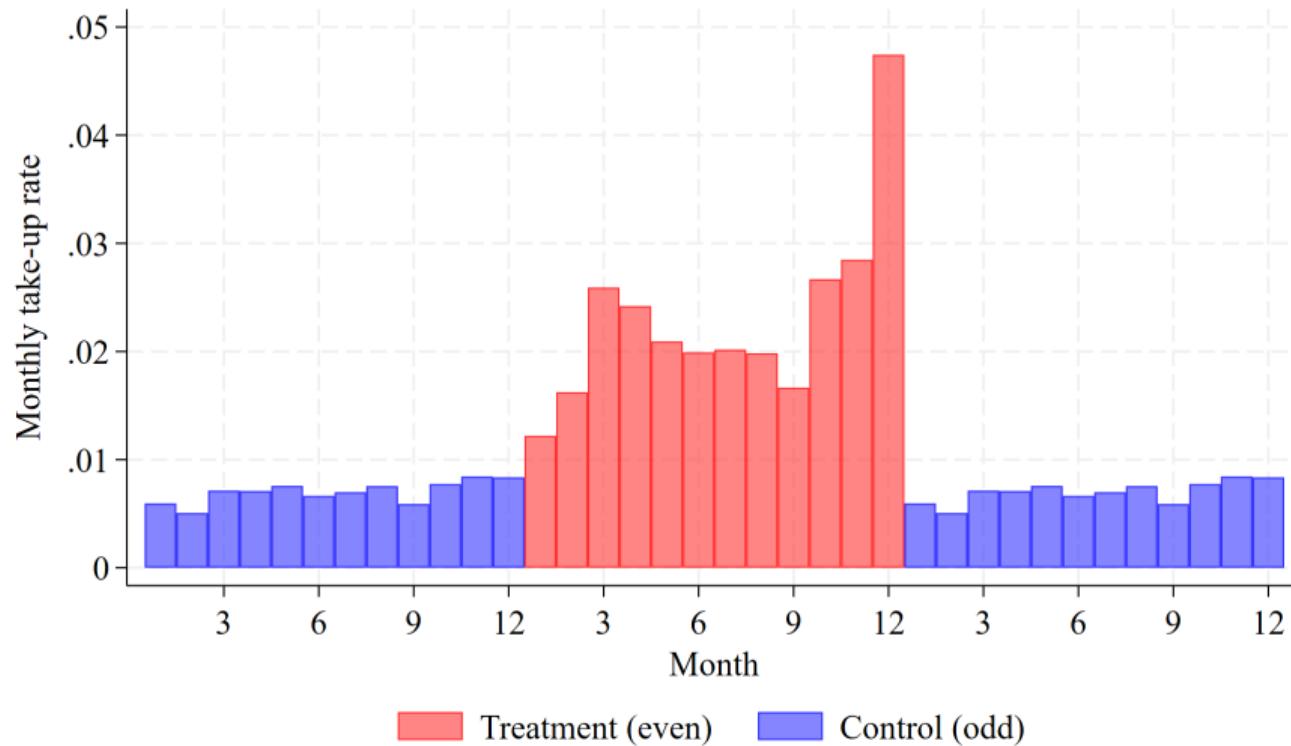
## No substitution for participants at age 38



## No substitution for participants at age 39



# People get screening earlier before subsidies expire



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## Multiple screenings on the same day

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Stomach	Breast	Cervical	Liver	Colorectal	Prostate	Lung
$\Pr(\text{general} = 1 \mid \text{screen} = 1)$	0.878	0.861	0.834	0.844	0.799	0.786	0.699
$\Pr(\text{same day} \mid \text{screen} = 1, \text{general} = 1)$	0.964	0.947	0.899	0.948	0.856	0.960	0.937
If not on the same day							
$\Pr(\text{general first} \mid \text{screen} = 1, \text{general} = 1)$	0.028	0.037	0.053	0.035	0.111	0.024	0.043
$\Pr(\text{general later} \mid \text{screen} = 1, \text{general} = 1)$	0.007	0.015	0.032	0.009	0.178	0.004	0.005

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# Women do not show stronger cross spillover

	(1)	(2)	(3)
	Liver	Colorectal	Lung
Age even	0.025*** (0.002)	0.036*** (0.002)	0.007*** (0.001)
Age even × Female	0.003 (0.003)	-0.005* (0.003)	-0.002 (0.001)
Female	-0.017*** (0.002)	-0.012*** (0.002)	-0.0078*** (0.0009)
N	107183	107183	107183
Control group mean	0.028	0.027	0.009
Age range	[40, 89]	[40, 89]	[40, 89]

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# Screening results

- Screening results

- Find any disease? ⇒ Which disease? (ICD-10)
- Multiple answers allowed
- Not available for never-takers

Screening	Take-up	Disease diagnosis
Stomach	17.9%	22.8%
Breast	16.4%	2.2%
Cervical	13.9%	6.7%
Colorectal	4.3%	22.3%

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## Disease classifications for stomach

- (K29) Gastritis and duodenitis
- (K52) Other noninfective gastroenteritis and colitis
- (K21) Gastro-oesophageal reflux disease
- (K25) Gastric ulcer
- (B98) Helicobacter pylori
- (K31) Other diseases of stomach and duodenum
- (K20) Esophagitis
- (C16) Malignant neoplasm of stomach
- (K26) Duodenal ulcer

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## Disease classifications for breast

- (N63) Unspecified lump in breast
- (N64) Other disorders of breast
- (D24) Benign neoplasm of breast
- (N60) Benign mammary dysplasia
- (C50) Malignant neoplasm of breast

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## Disease classifications for female genital

- (N76) Other inflammation of vagina and vulva
- (N71) Inflammatory disease of uterus, except cervix
- (N85) Other noninflammatory disorders of uterus, except cervix
- (N83) Noninflammatory disorders of ovary, fallopian tube and broad ligament

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## Disease classifications for colon and rectum

- (K63) Other diseases of intestine
- (D12) Benign neoplasm of colon, rectum, anus and anal canal
- (D13) Benign neoplasm of other and ill-defined parts of digestive system
- (R19) Other symptoms and signs involving the digestive system and abdomen
- (C18) Malignant neoplasm of colon

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# 1. Estimate Always- and Never-takers characteristics

- Individually identifiable always- and never-takers

	Even age (treatment)	Odd age (control)
Always-takers	Yes	Yes
Compliers	Yes	No
Never-takers	No	No

- Estimating equation

$$y_{it} = \beta_0 + \beta_1 treat_{it} + \beta_2 screen_{it} + \beta_3 treat_{it} \times screen_{it} + \nu_{it} \quad (3)$$

- Average characteristics

- Always-takers:  $g_{AT}(y) = \hat{\beta}_0 + \hat{\beta}_2$
- Never-takers:  $g_{NT}(y) = \hat{\beta}_0 + \hat{\beta}_1$

## 2. Back out complier characteristics

- Treated compliers in the treatment group, untreated compliers in the control group

	Even age (treatment)	Odd age (control)
Always-takers	Yes	Yes
Compliers	Yes	No
Never-takers	No	No

- Estimating equation

$$y_{it} = \beta_0 + \beta_1 treat_{it} + \beta_2 screen_{it} + \beta_3 treat_{it} \times screen_{it} + \nu_{it} \quad (4)$$

- Those getting screened in the treatment group

$$\begin{aligned} g_T(y) &= \frac{\pi_{AT}}{\pi_{AT} + \pi_C} g_{AT}(y) + \frac{\pi_C}{\pi_{AT} + \pi_C} g_C^1(y) \\ &= \hat{\beta}_0 + \hat{\beta}_1 + \hat{\beta}_2 + \hat{\beta}_3 \end{aligned}$$

- Those not getting screened in the control group

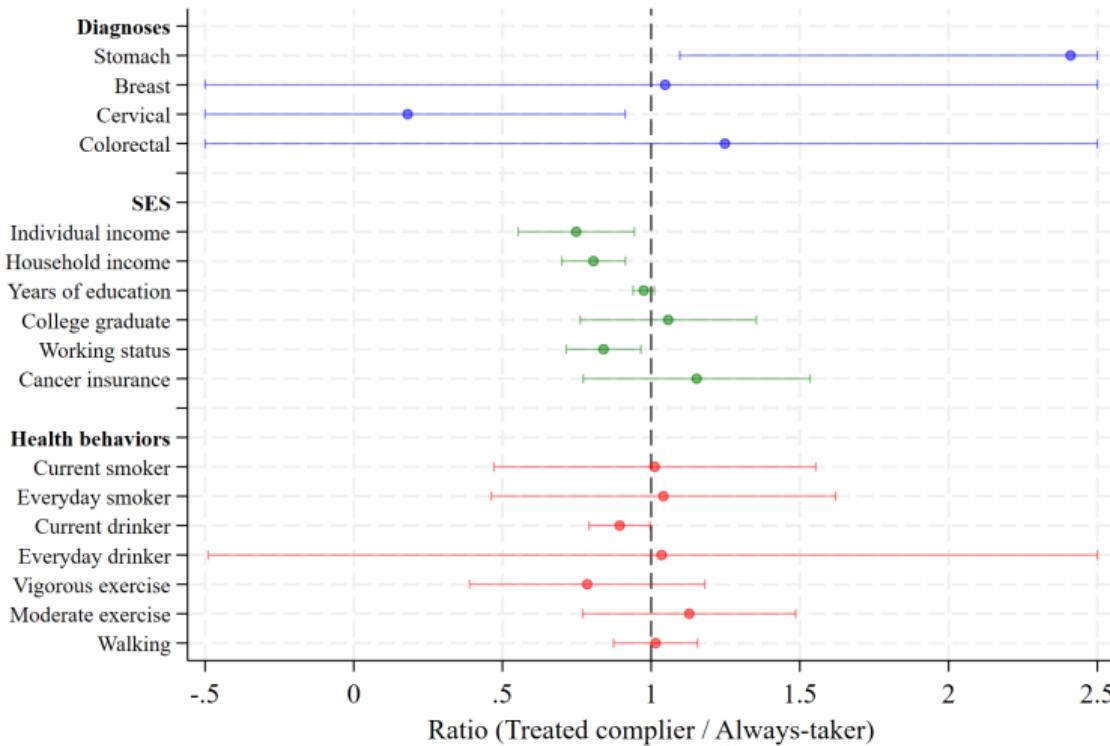
$$\begin{aligned} g_U(y) &= \frac{\pi_{NT}}{\pi_{NT} + \pi_C} g_{NT}(y) + \frac{\pi_C}{\pi_{NT} + \pi_C} g_C^0(y) \\ &= \hat{\beta}_0 \end{aligned}$$

### 3. Compare compliers to always- and never-takers

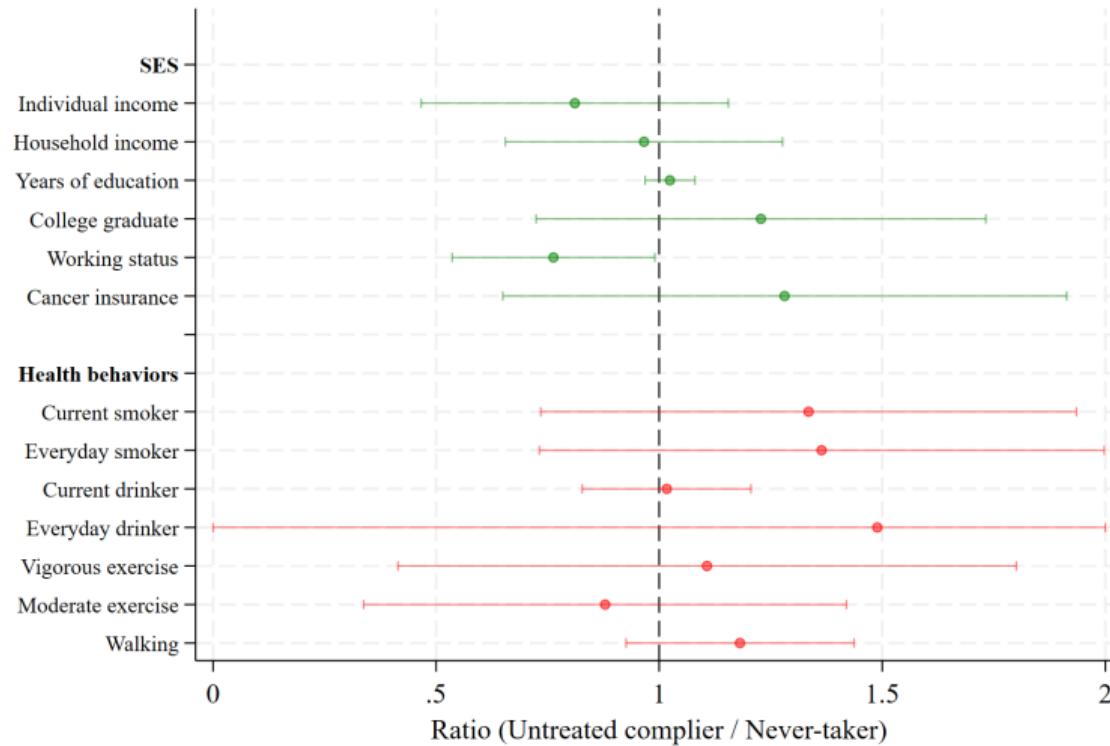
- Taking ratios
  - Treated compliers to always-takers:  $g_C^1(y)/g_{AT}(y)$
  - Untreated compliers to never-takers:  $g_C^0(y)/g_{NT}(y)$
- Why differentiate between treated and untreated compliers?
  - Characteristics in the same year
  - Unclear pre-determined characteristics
  - Difference between treated and untreated complier characteristics = LATE
- Minor details in estimation
  - Age = 60
  - Standard error calculated with bootstrap with clustering at individual level (500 replications)

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# Compliers vs Always-takers using age 40 discontinuity



# Compliers vs Never-takers using age 40 discontinuity



# Stomach diseases and cancer

- Unlike in US, stomach cancer is common in South Korea
  - % of all new cancer cases: 1.3% (US 2024) 11.5% (SK 2019)
  - % of all cancer deaths: 1.8% (US 2024) 9.4% (SK 2019)
- There is high returns to early detection of stomach cancer
  - 5-year relative survival rate
  - Stomach: localized (97.4%) → distant (6.6%)
  - Breast: localized (98.9%) → distant (45.2%)
- Stomach screening (upper endoscopy)
  - Cause: Helicobacter Pylori ([Uemura et al., 2001](#))
  - Symptoms: inflammation (gastritis), ulcer (peptic ulcer), cancer
  - Diagnosed through endoscopy
  - Treatment: 2 weeks of antibiotics and an acid-reducing proton pump inhibitor

# Stomach diseases and cancer

	(1)	(2)	(3)
	Average value		Ratio
	Always-takers	Treated compliers	$CP_1 / AT$
Helicobacter Pylori	0.009 (0.002)	0.013 (0.002)	1.498 (0.360)
Gastritis	0.123 (0.005)	0.210 (0.006)	1.707*** (0.101)
Stomach ulcer	0.005 (0.001)	0.008 (0.001)	1.444 (0.529)
Stomach cancer	0.001 (0.001)	0.003 (0.001)	1.909 (1.924)
Colon polyp	0.070 (0.007)	0.114 (0.016)	1.628* (0.359)

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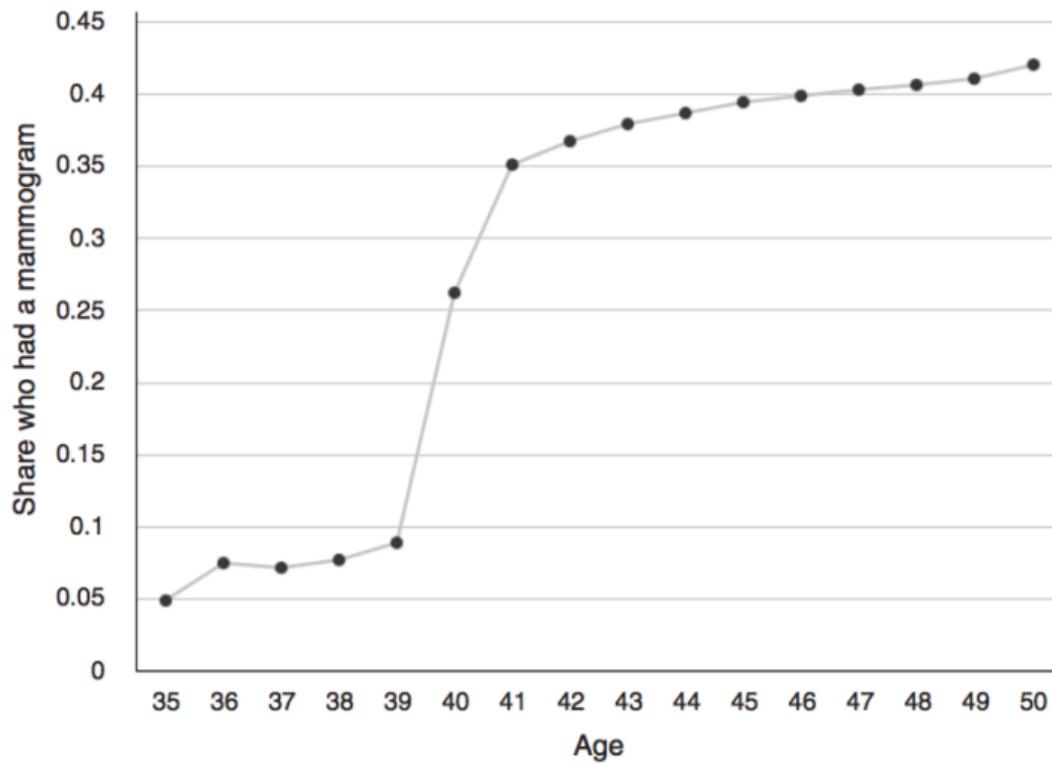
# Using 10 years of panel to define compliance groups

	Treatment (even)	Control (odd)
Always-takers	Yes	Yes
Compliers	Yes	No
Defiers	No	Yes
Never-takers	No	No

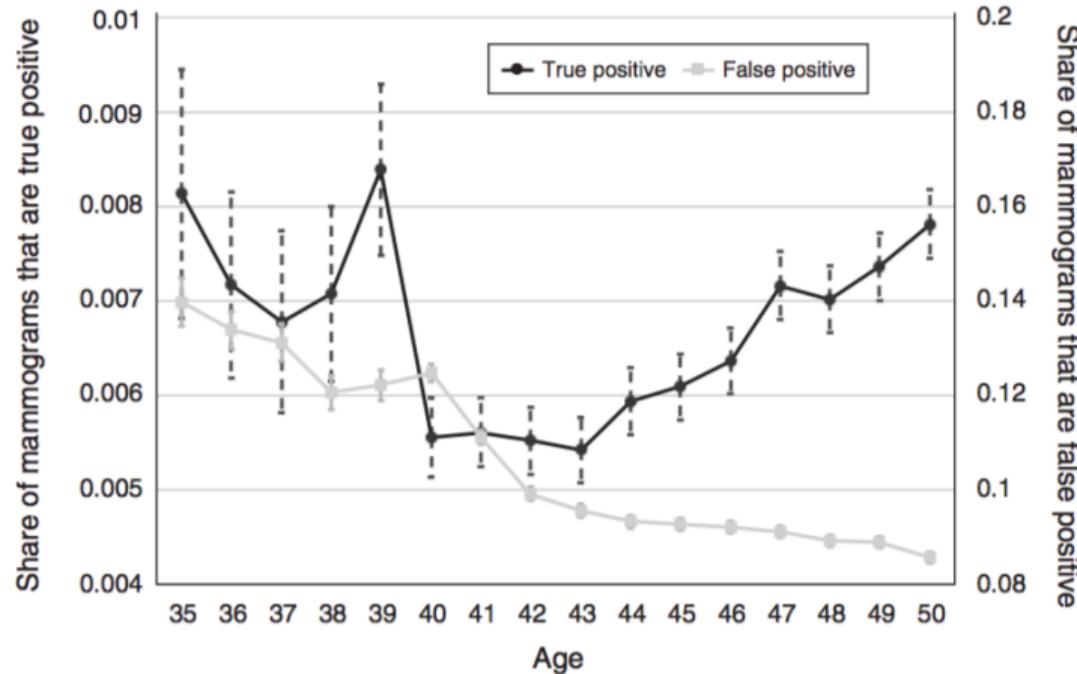
- Analytical sample ( $N = 5,701$ )
  - Balanced sample without attrition for 10 years (2009-2018)
  - First year age  $\geq 40$
- Probability of screening at even and odd ages

$$Pr(\text{screen\_even}) = \frac{1}{5} \times \sum_{k=1}^5 \mathbb{1}\{\text{screen}_{ik} = 1\}, \quad k \text{ even} \quad (5)$$

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Panel A. Share of mammograms that are true positive and false positive



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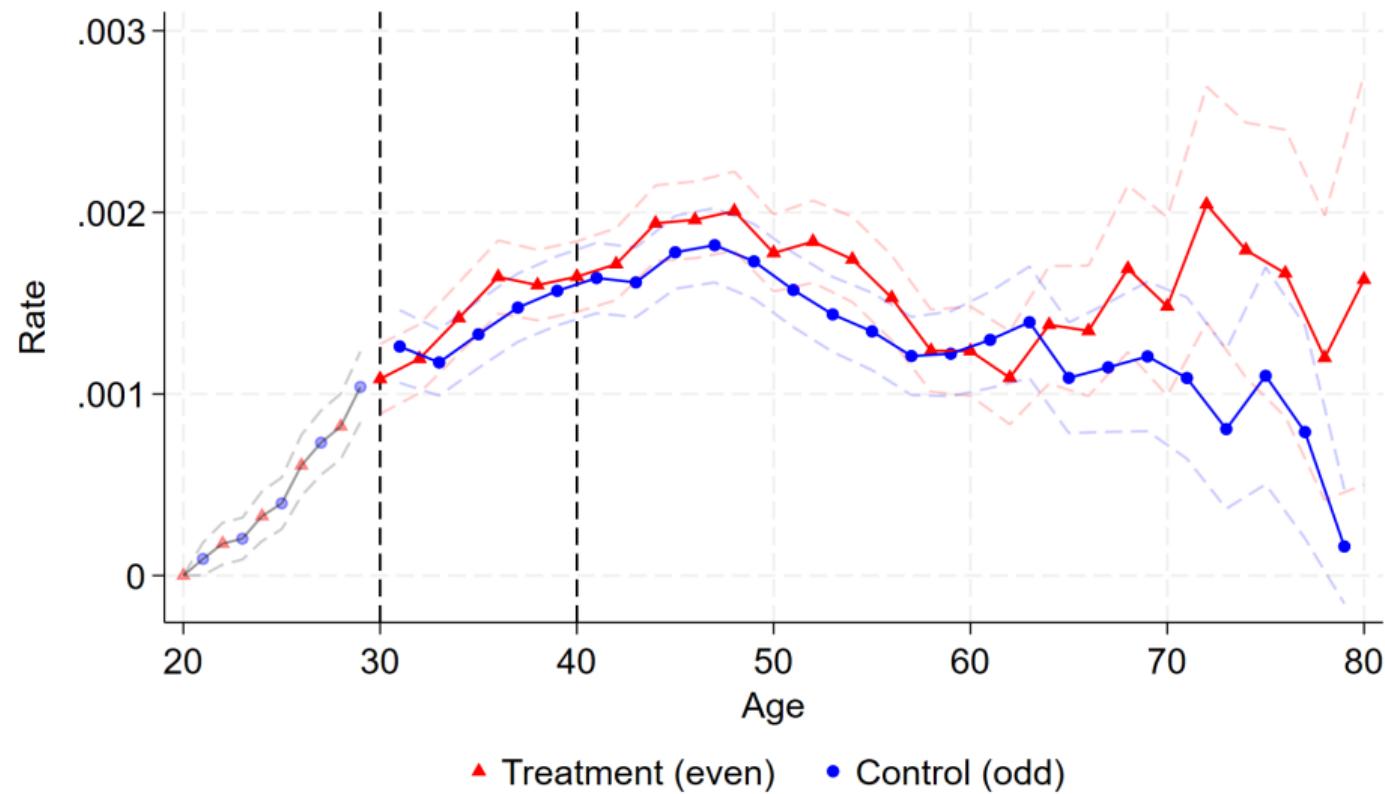
    Effect on take-up

    Cross spillover

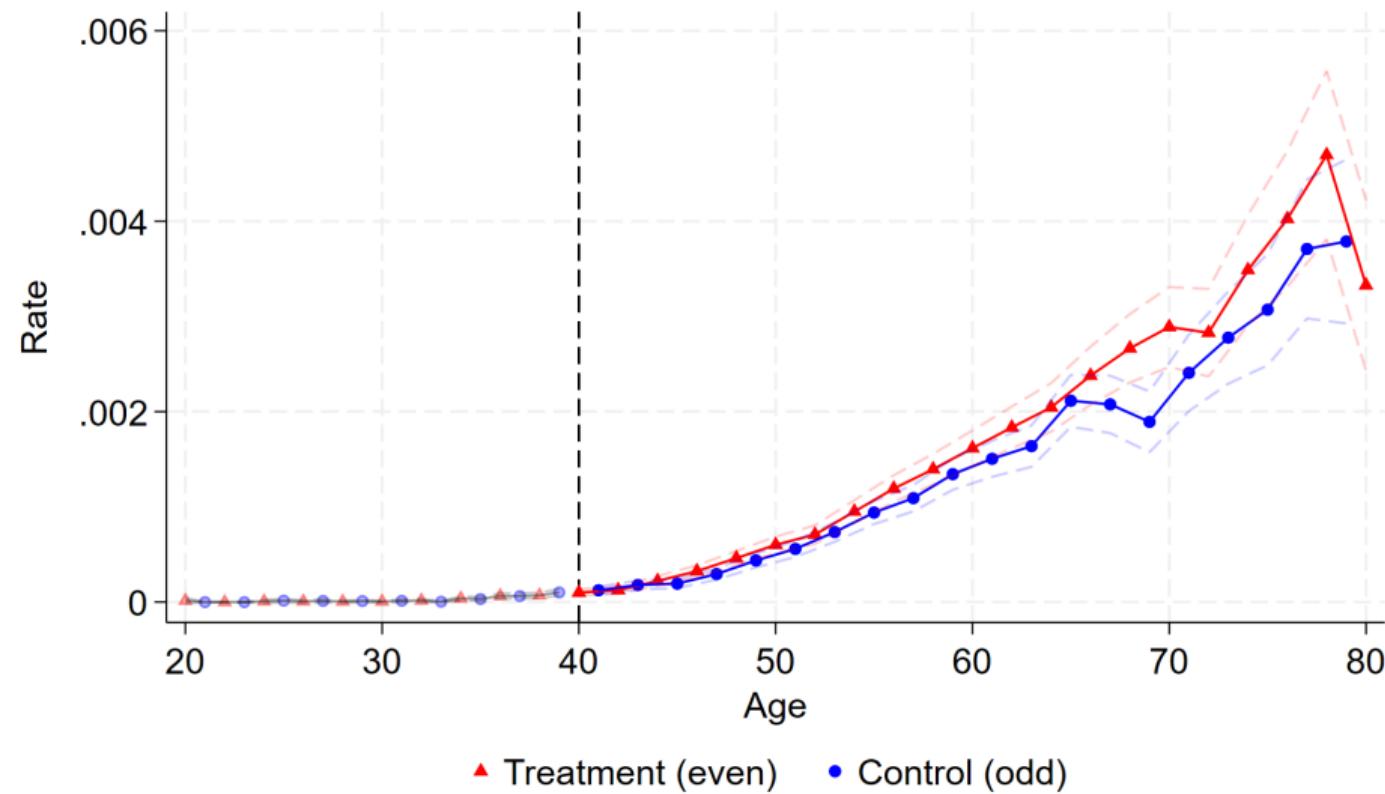
    Selection

**Effect**

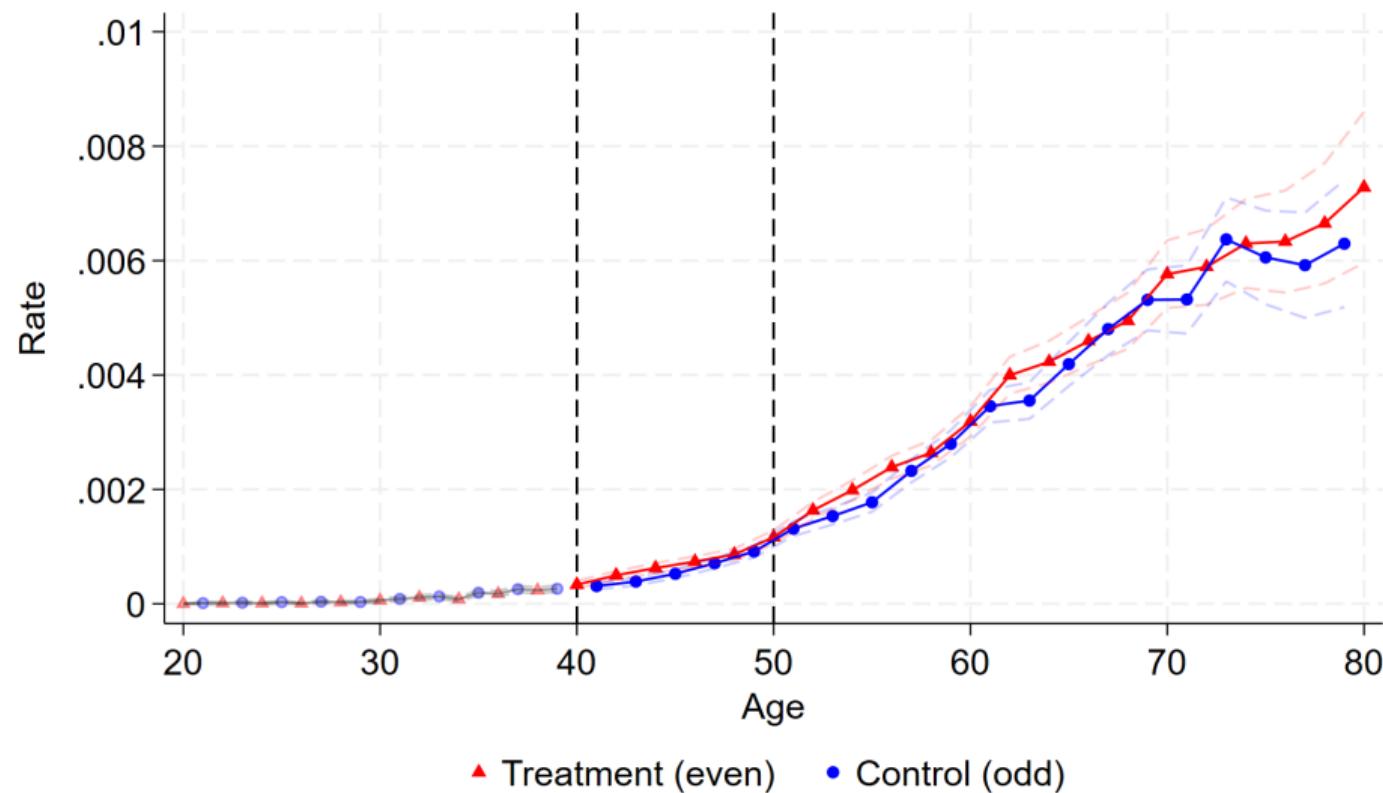
## More cervical cancer diagnoses in the treatment group



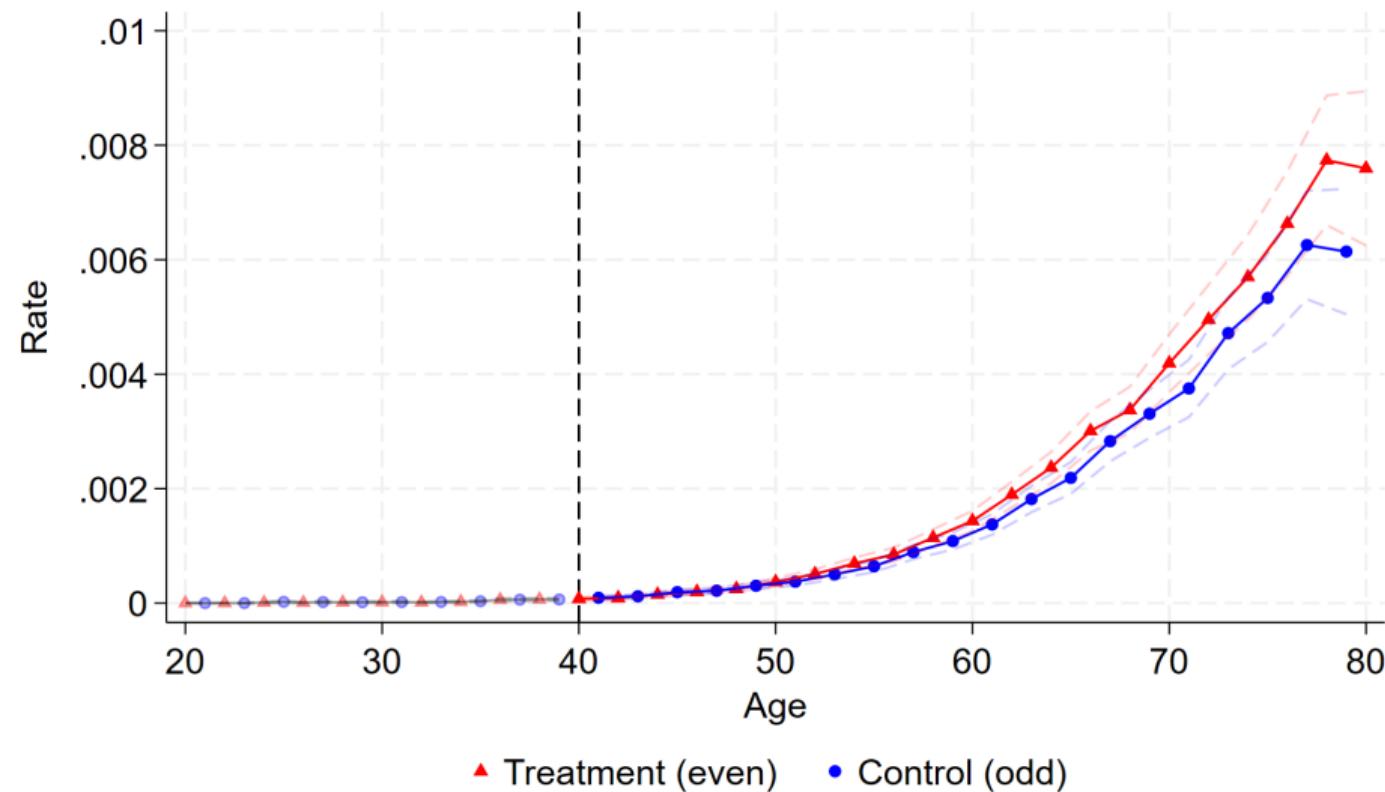
## More liver cancer diagnoses in the treatment group



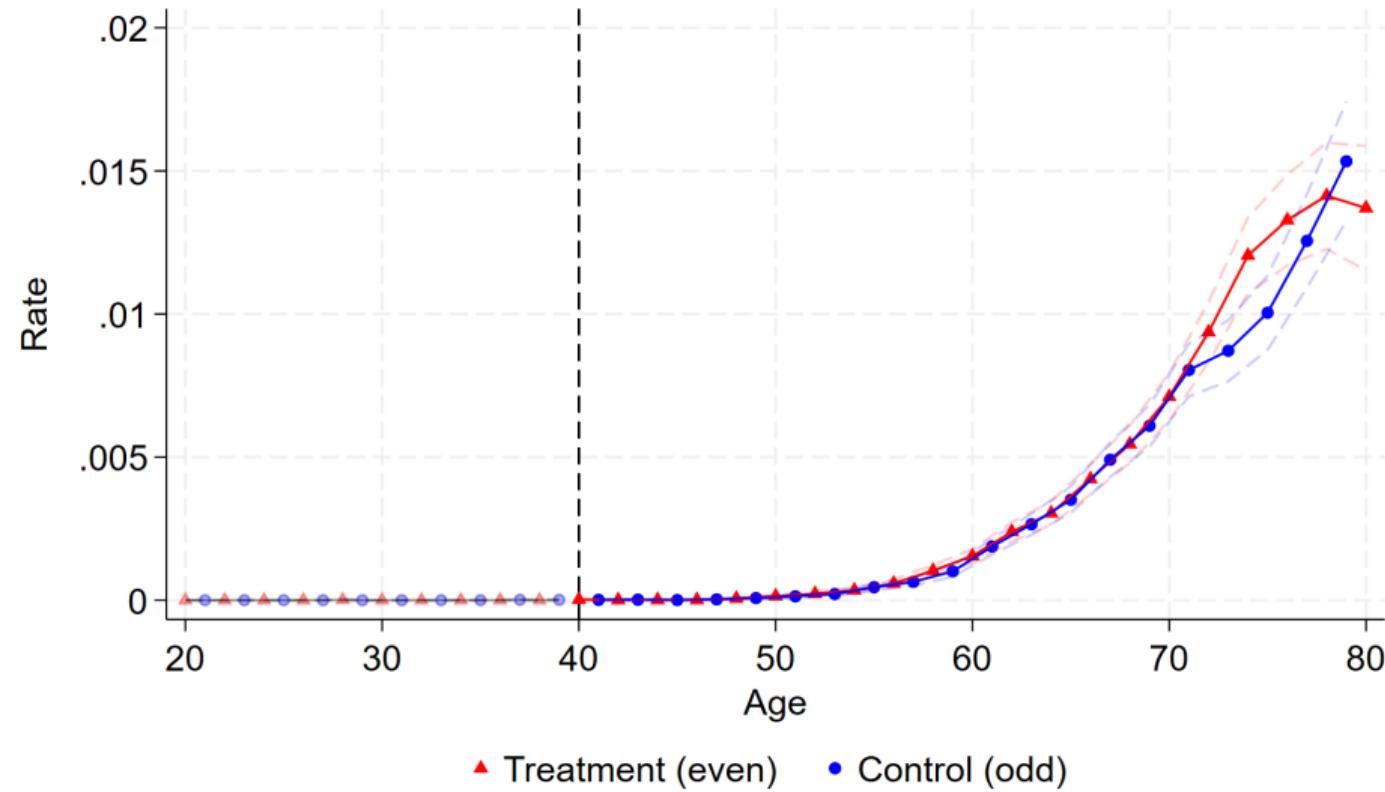
## More colorectal cancer diagnoses in the treatment group



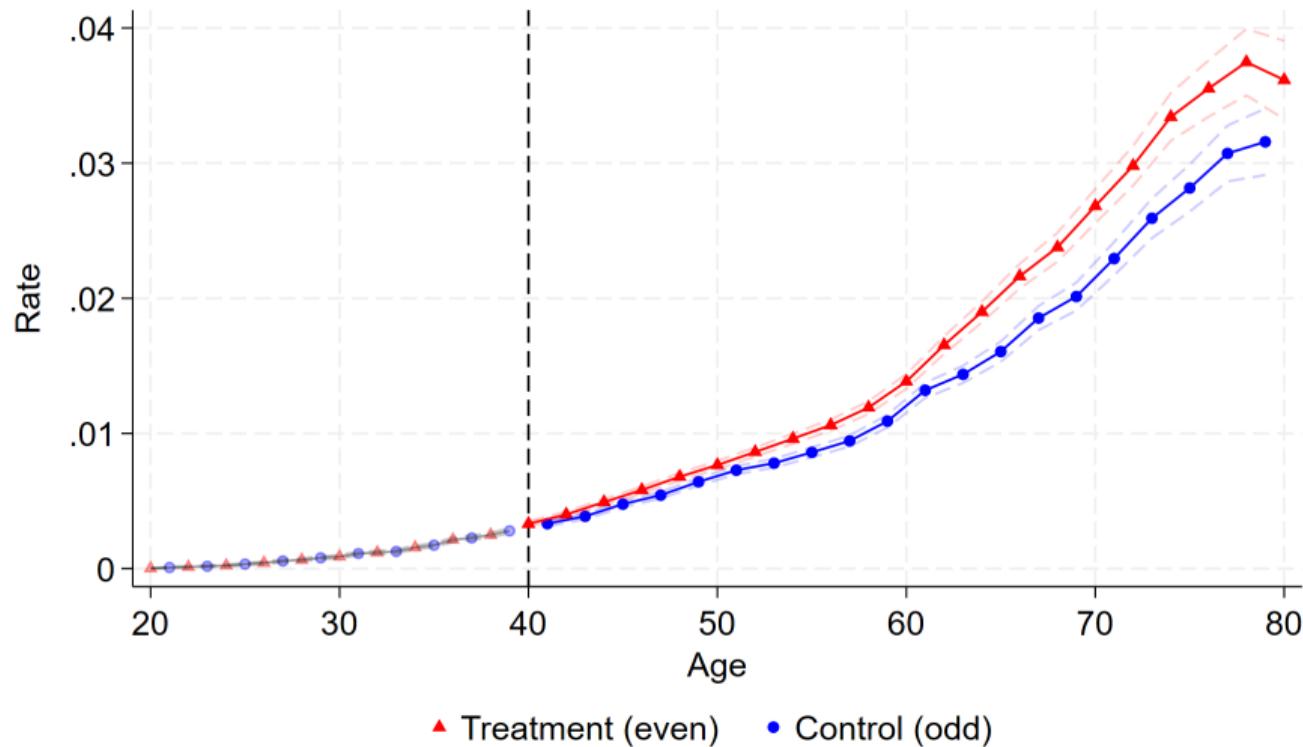
## More lung cancer diagnoses in the treatment group



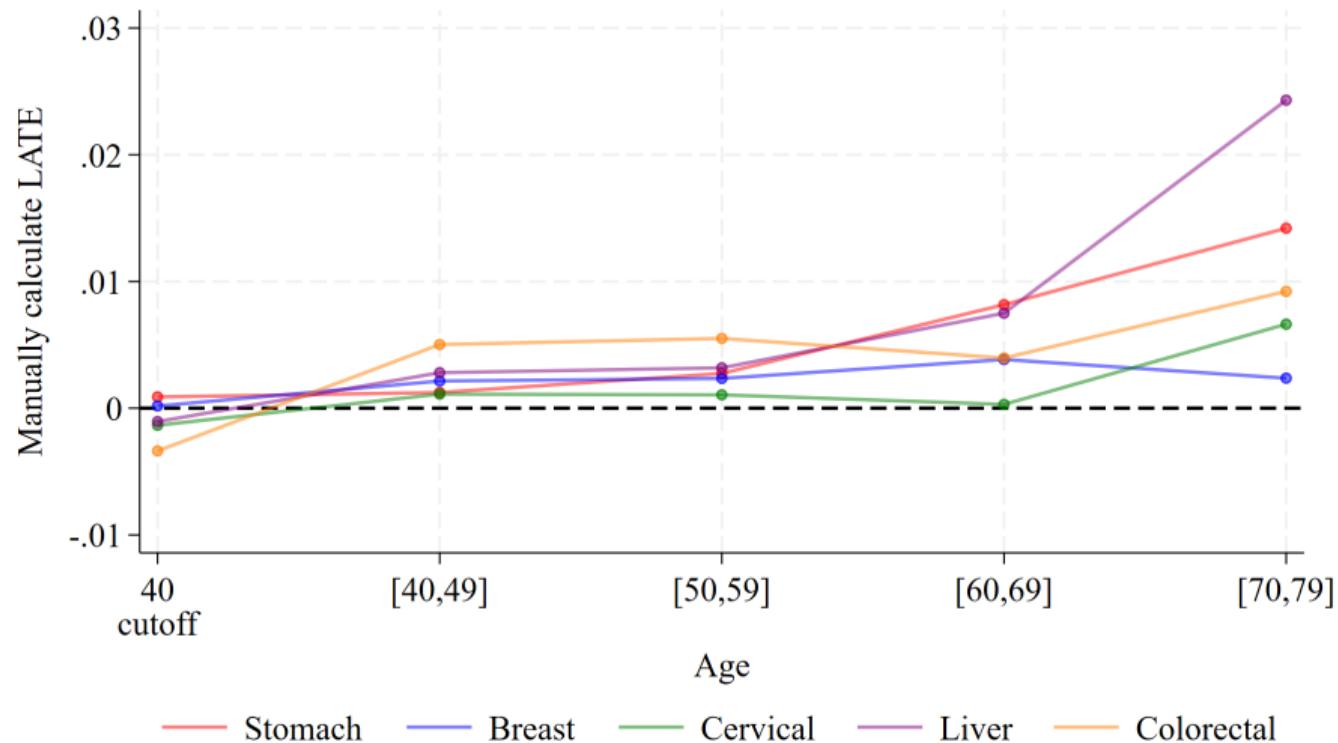
## 'Slightly' more prostate cancer diagnoses in the treatment group



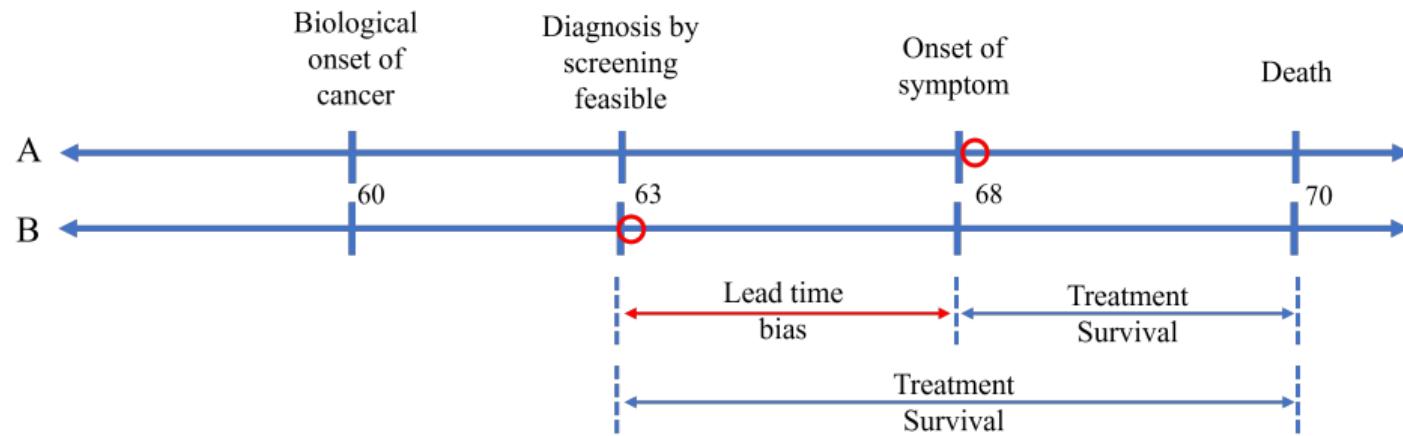
## Overall, screening leads to more cancer diagnoses



## Manually calculated LATE estimates by age



# Lead time bias in survival rate analysis



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# Screening increases medication use for high blood pressure

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
High blood pressure	0.23884	0.00046*** (0.00013)	0.193	8,673,954
ARBs (Angiotensin II Receptor Blockers)	0.10179	0.00038*** (0.00009)	0.378	8,673,954
Calcium Channel Blockers (CCBs)	0.09779	0.00021** (0.00009)	0.218	8,673,954
Diuretics (Water Pills)	0.07136	-0.00003 (0.00010)	-0.046	8,673,954
Beta-Blockers	0.04289	0.00005 (0.00008)	0.121	8,673,954
ACE Inhibitors	0.01357	-0.00003 (0.00004)	-0.233	8,673,954
Aldosterone Antagonists	0.00970	-0.00007 (0.00004)	-0.705	8,673,954
Alpha-Blockers	0.00768	0.00001 (0.00004)	0.109	8,673,954

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# Screening increases medication use for diabetes

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
Diabetes	0.08918	0.00015** (0.00006)	0.171	8,673,954
Biguanides	0.06832	0.00011* (0.00007)	0.165	8,673,954
Sulfonylureas	0.04318	-0.00006 (0.00006)	-0.139	8,673,954
DPP-4 Inhibitors	0.01836	0.00005 (0.00004)	0.298	8,673,954
Alpha-Glucosidase Inhibitors	0.00432	0.00001 (0.00002)	0.195	8,673,954
Meglitinides	0.00174	-0.00001 (0.00002)	-0.303	8,673,954
Insulin (Various Types)	0.00165	0.00001 (0.00001)	0.615	8,673,954
GLP-1 Receptor Agonists	0.00014	0.00000 (0.00001)	3.127	8,673,954
SGLT2 Inhibitors	0.00202	-0.00000 (0.00002)	-0.204	8,673,954
Thiazolidinediones (TZDs)	0.00671	-0.00002 (0.00003)	-0.299	8,673,954

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# Screening increases medication use for high cholesterol

	(1)	(2)	(3)	(4)
	Control group mean	ITT	Percent relative to control	N
High cholesterol	0.13901	0.00419*** (0.00011)	3.014	8,673,954
Statins	0.13359	0.00405*** (0.00011)	3.033	8,673,954
Fibrates	0.00752	0.00016*** (0.00004)	2.076	8,673,954
Ezetimibe	0.00194	0.00006*** (0.00002)	3.131	8,673,954

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