Health and Economic Growth

Econ 30250 Bill Evans Spring 2020

Preston Curve

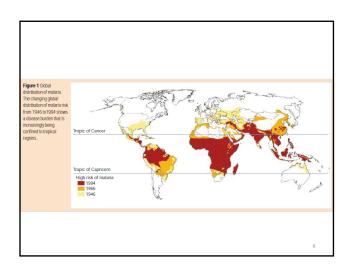
- Suggestive of a causal link greater economic success increases life expectancy
- Could also suggest health is key to development
 economies grow with a healthy population
- Belief by many that poor health is holding back the development of many countries – especially in Africa

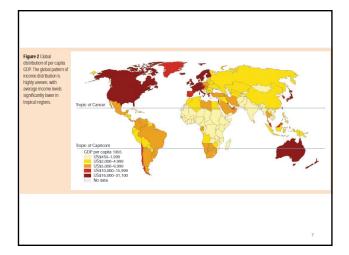
Many interesting questions

- Role of rising incomes?
- What do those rising incomes purchase that allows mortality to fall?
- Can you "jump start" the change in mortality?
- Q we are going to consider is a little different does health detract from growth and can a healthier population improve economic returns?
 - Some suggestive evidence from previous section

Case study: Malaria

- Burden
 - 300-500 million cases per year
 - 1 -3 million fatalities, mostly children
 - 90% of malaria mortality in Africa
- · Centered on tropics
 - Transmission less likely when temp <18°C (64.4)
 - Parasite dies at 16°C (60.8)
- Has been successfully eradicated in the US





Jeff Sachs

 "...malaria not only takes an enormous human toll in Africa, but also contributes to an enormous economic loss and is a barrier to economic growth. Investments in malaria control thus offer an enormous return in lives saved and in economic benefits for Africa."

What we do in this lecture

- Isolate pathways through which health can impact growth. Provide:
 - Theoretical link
 - Empirical evidence for each of these links
 - Emphasis on historical data
- Some examples rapid changes in mortality does it impact health?

Bloom and Canning

4 pathways linking health to growth

- 1. Productivity
- 2. Education
- 3. Investments in Physical Capita
- 4. Demographic dividend

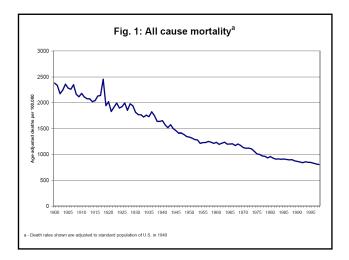
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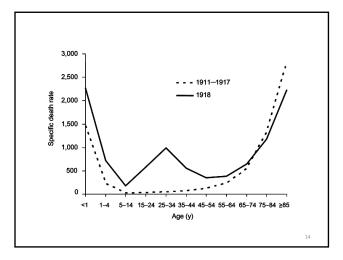
Health and productivity

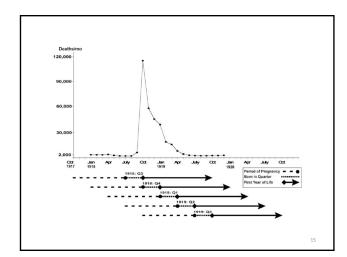
- Many good papers demonstrate a link between heath shocks and productivity later in life
- Much from developing country
- One quick example from the US -- 1918 Flu epidemic

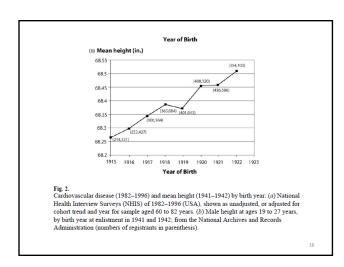
1918 Flu Epidemic

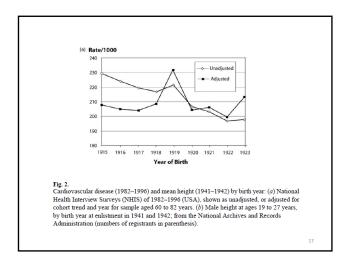
- Spanish flu
- World wide epidemic
 - Killed 30-50 million, 675K in the US
- Those particularly vulnerable
 - Children
 - Compromised immune system
 - Pregnant women

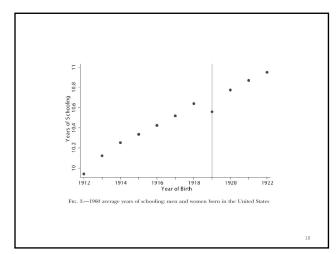












Bloom and Canning

4 pathways linking health to growth

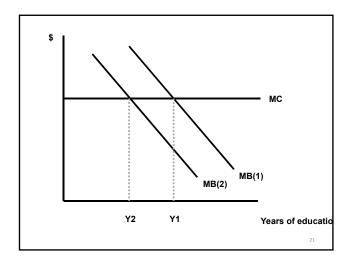
- 1. Productivity
- 2. Education
- 3. Investments in Physical Capital
- 4. Demographic dividend

Income C D

\$0

A Invest of PV(A+B)<PV(C)
Invest of PV(A+B)<PV(C+D)

18 22 55 65



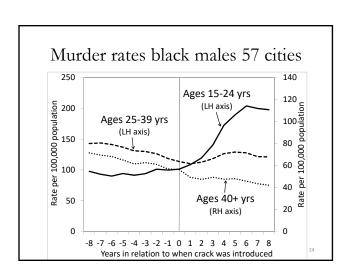
Evidence: Rise of Crack Cocaine

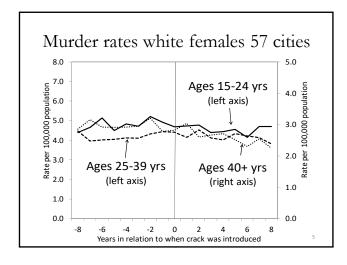
- Crack enters in 1982 on coasts spreads to the center of the country
- Devastating to young black males
 - $-2x\uparrow$ murder rate
 - $-4x \uparrow$ in incarceration rates
- Human capital models should see \(\) investment
 - − ↓ life expectancy
 - – ↓ job prospects (due to prison records)
 - ↑ "outside" option

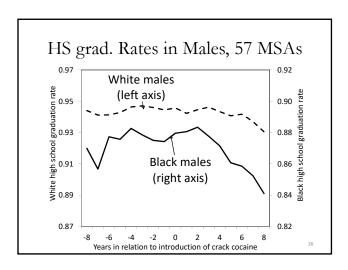
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When Crack Arrives

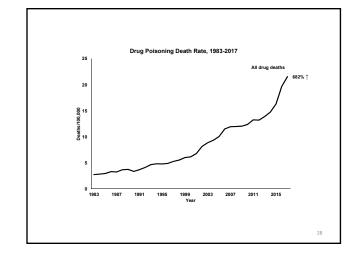
- 1982: NY, LA, Miami
- 1983: Atlanta, Riverside, SF
- 1984: Seattle, Tampa, San Jose, Ft. Lauderdale
- 1985: Detroit, Houston, KC, Orange Co., Philly, DC
- 1986: Boston, Chicago, Cleveland, Indy, Memphis, MSP, New Orleans, Newark, Sacramento
- 1987: Dallas, Portland, Milwaukee, Hartford, Newark, Providence, Greesnboro/WS

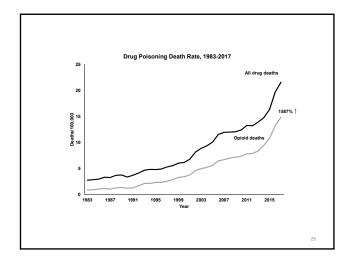


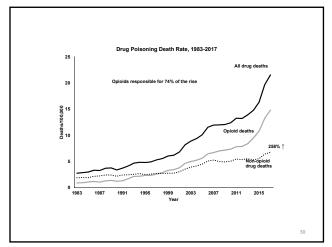




Are there other situations where there are rapid changes in mortality that one can use in the same manor?







Some facts

- Drug deaths
 - **-** 1983: 6,445
 - **2017: 72,000**
 - M&T Bank Stadium: 71,008
- Peak deaths/year for other epidemics
 - HIV/AIDS: 41,669 in 1995
 - Motor vehicle fatalities: 51,903 in 1978
 - Murder: 24,530 in 1993

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Some facts

- 704,000 drug poisoning deaths between 1999 2017
 - Military deaths Spanish American War through now -- 702,000
- 400,000 opioid deaths since 1999
 - -# of US deaths in WW II

Table 1. All-Cause Mortality for White Non-Hispanics with High School or Less and All Black Non-Hispanics by Five-Year Age Cohort, 1999 and 2015

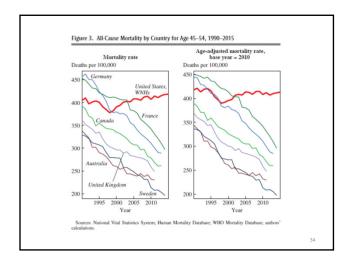
| | 1999 | | 2015 | | | |
|-------|---|-------------|---|-------------|--|--|
| Age | White non-Hispanics, high school or less | Blacks, all | White non-Hispanics, high school or less | Blacks, all | | |
| 25-29 | 145.7 | 169.8 | 266.2 | 154.6 | | |
| 30-34 | 176.8 | 212.0 | 335.5 | 185.5 | | |
| 35-39 | 228.8 | 301.4 | 362.8 | 233.6 | | |
| 40-44 | 332.2 | 457.4 | 471.4 | 307.2 | | |
| 45-49 | 491.2 | 681.6 | 620.1 | 446.6 | | |
| 50-54 | 722.0 | 945.4 | 927.4 | 703.1 | | |
| 55-59 | 1,087.6 | 1,422.8 | 1,328.3 | 1,078.9 | | |
| 60-64 | 1,558.4 | 1,998.3 | 1,784.6 | 1,571.1 | | |

Sources: National Vital Statistics System; authors' calculations.

White 25-29: 82% increase Black 25-29: 8% decline

White 30-34: 90% increase Black 30-34: 12.5% decline

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Bloom and Canning

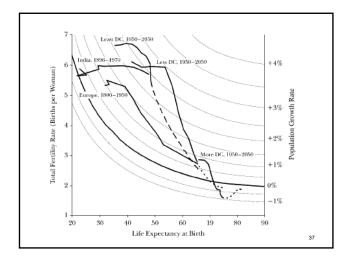
4 pathways linking health to growth

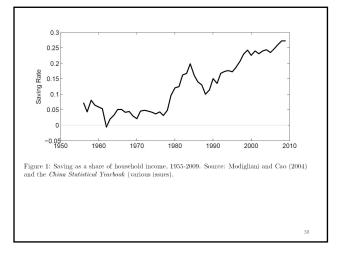
- 1. Productivity
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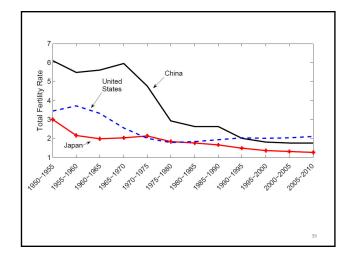
Bloom and Canning

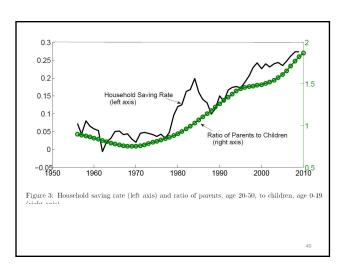
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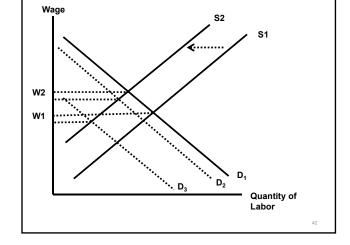






Alternate hypothesis: Black Plague

- Plague strikes Europe 1348-1350
- Carried by flees living on black rats
- Shipping routes spread the disease quickly
- Kills 75 200 million
- Reduces pre-plague population in England by 50%

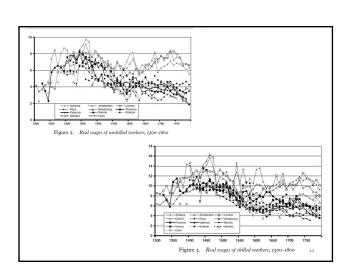


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Table I. Population of selected European countries, 1300–1800 (in thousands)

| | 1300 | 1400 | 1500 | 1600 | 1700 | 1800 |
|-------------------|---------|--------|--------|---------|---------|---------|
| England and Wales | 5,750 | 3,000 | 3,500 | 4,450 | 5,450 | 9,250 |
| Netherlands | 800 | 600 | 950 | 1,500 | 1,950 | 2, 100 |
| Belgium | 1,250 | 1,000 | 1,400 | 1,600 | 2,000 | 2,900 |
| Italy | 12,500 | 8,000 | 9,000 | 13,300 | 13,500 | 18, 100 |
| Spain | 5,500 | 4,500 | 5,000 | 6,800 | 7,400 | 11,000 |
| Total Europe | 94, 200 | 67,950 | 82,950 | 107,350 | 114,950 | 192,230 |

Source: Paolo Malanima (unpublished manuscript).



Consequences

- Europe in 1300s was mired in stagnant wages and high population
- Massive decline in population increased value of labor
- Jump-started income growth in Europe
- Young: "Gift of Dying." Argues the same for Africa and AIDS

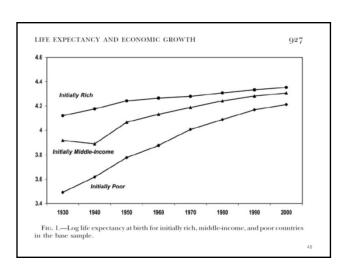
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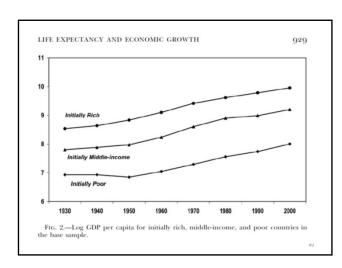
Acemoglu and Johnson (JPE)

- International epidemiological transition
 - Began in 1940
 - Large improvements world wide in life expectancy
- Three factors
 - Drugs (mass production of penicillin, antibiotics), vaccines (polio, measles, etc.), DDT
 - WHC
 - Change in universal values encouraged spread of changes to poor countries

- IDT was "technology" based
- Therefore it impacted poor countries the most (impacted those most in need)
- Exogenous change in mortality
- Since it impacted poor countries the most, we should see a greater change in GDP for this group if health has an impact on the economy







Explaining results

- Drop in mortality increases population
- Should increase output
- BUT -- because capital is fixed
 - Capital used more intensely
 - Productivity declines, reduces wages
- Growth in output from more people is not enough to compensate for loss in productivity per worker
- Black plague argument

| | WHOLE WORLD | Base S | Sample | Low- and Middle- Income Countries Only | Base Sample | LOW- AND MIDDLE- INCOME COUNTRIES ONLY |
|---------------------|-----------------------|-----------------------|-----------------------|--|-----------------------|--|
| | (1) | (2) | (3) | (4) | (5) | (6) |
| | | A. De | ependent Var | riable: Log Pop | oulation | |
| | Just 1960 and 2000 | Just 1960 and 2000 | Just 1940 and 1980 | Just 1940 and 1980 | Just 1940 and 2000 | Just 1940 and 2000 |
| Log life expectancy | 1.60 (.30) | 1.75 (.40) | 1.62 (.19) | 1.86 (.26) | 2.01 (.22) | 2.25 (.32) |
| Number of countries | 120 | 59 | 47 | 36 | 47 | 36 |

| | | | STIMATES | | | | |
|--|---|------------------------------|------------------------------|---|------------------------------|---|--|
| | WHOLE WORLD: Just 1960 and 2000 (1) | Base Sample | | LOW- AND MIDDLE- INCOME COUNTRIES ONLY: | BASE SAMPLE: | LOW- AND MIDDLE- INCOME COUNTRIES ONLY: | |
| | | Just 1960 and 2000 (2) | Just 1940 and 1980 (3) | Just 1940 and 1980 (4) | Just 1940 and 2000 (5) | Just 1940 and 2000 (6) | |
| | A. Dependent Variable: Log GDP | | | | | | |
| Log life expectancy Number of countries | 1.17 (.56) 120 | 1.55 (.35) 59 | .78 (.33) 47 | .65 (.42) 36 | .85 (.28) 47 | .43 (.38) 36 | |
| Number of countries | 120 | | | ole: Log GDP 1 | | 50 | |
| Log life expectancy | 42 (.58) | 19 (.54) | 81 (.26) | -1.17 (.38) | -1.14 (.27) | -1.79 (.41) | |
| Number of countries | 120 | 59 | 47 | 36 | 47 | 36 | |

Cutler et al., Malaria Eradication in India

- Will reductions in Malaria necessarily lead to higher education?
- What are definitive predictions about outcomes?
 - Income/consumption
 - education?

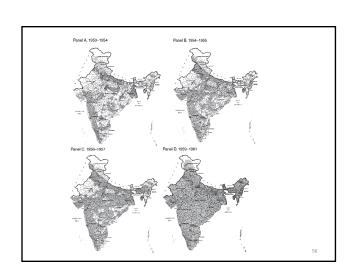
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Malaria Eradication in India

- National Malaria Control Program launched April of 1953
- · Heavy use of DDT
 - Effective, nontoxic for humans, cheap
 - Eradicated malaria in Taiwan, Caribbean, Balkans, parts of North Africa, north Australia, large parts of South Pacific
- Prior to program, 75 million annual cases in India and 800K annual deaths (~350 million people)

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- Two annual rounds of spraying
 - 1/3 of country initially part of program
 - Program reformulated in 1958
 - Whole country part of program in 1960-61
- Strategy Difference-in-Difference
 - Compare outcomes of groups some born before and after eradication program
 - Variation in timing of program across regions
 - Some areas had higher pre-treatment malaria rates so allow treatment to vary



Model

 $y_{icd} = x_{icd}\gamma + POST_cxMalaria_d\beta + \delta_d + \alpha_c + \varepsilon_{icd}$

i = person, c = cohort, d = district

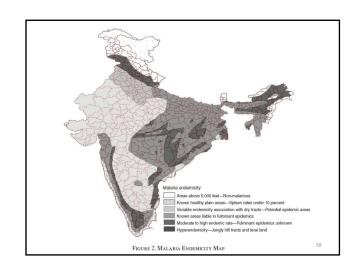
y = outcome

x = cov ariates

 $POST_c = 1$ if cohort was born after eradication program $Malaria_d = malaria$ incidence rate prior to program

 $\alpha_d = district \ effects$

 $\delta_c = cohort \ effects$



| | Literacy (ages 15–75) | | | | Primary school (ages 15–75) | | | |
|-----------------------------------|--------------------------|------------------|-------------------|------------------|--------------------------------|------------------|-------------------|------------------|
| Dependent variable: | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) |
| A2. Districts classified by aver | age malaria cate | gory | | | | | | |
| $Post \times malaria \ index$ | -0.017 (0.006)*** | 0.004 (0.005) | -0.001 (0.006) | 0.008 (0.011) | -0.016 (0.007)** | 0.002 (0.006) | -0.005 (0.008) | 0.009 |
| Observations | 111,139 | 111,139 | 111,139 | 111,139 | 111,139 | 111,139 | 111,139 | 111,139 |
| State × post fixed effects | | X | | | | X | | |
| Region × post fixed effects | | | X | X | | | X | X |
| District-specific linear trends | | | | X | | | | X |
| B2. Districts classified by avera | ge malaria categ | gory | | | | | | |
| Post × malaria index | 0.005 (0.006) | 0.011 (0.006)* | -0.006 (0.006) | 0.008 (0.010) | -0.004 (0.007) | 0.005 (0.006) | -0.012 (0.008) | 0.002 (0.007) |
| Observations | 107,472 | 107,472 | 107,472 | 107,472 | 107,472 | 107,472 | 107,472 | 107,472 |
| State × post fixed effects | | X | | | | X | | |
| Region × post fixed effects | | | X | X | | | X | X |
| District-specific linear trends | | | | X | | | | X |

| | Log per capita household expenditure (ages 20-60) | | | | | | |
|--|---|--------------------|---------------------|------------------|--|--|--|
| Dependent variable: | (1) | (2) | (3) | (4) | | | |
| A2. Districts classified by average ma | laria category | | | | | | |
| Post × malaria index | 0.008 (0.004)** | 0.011 (0.005)** | 0.019 (0.006)*** | 0.008 (0.011) | | | |
| Observations | 75,230 | 75,230 | 75,230 | 75,230 | | | |
| State × post fixed effects Region×post fixed effects District-specific linear trends | | X | X | X X | | | |
| B2. Districts classified by average mal | aria category | | | | | | |
| Post × malaria index | -0.003 (0.004) | -0.003 (0.004) | 0.004 (0.005) | 0.011 (0.014) | | | |
| Observations | 75,212 | 75,212 | 75,212 | 75,212 | | | |
| State × post fixed effects Region × post fixed effects District-specific linear trends | | X | X | X X | | | |