The Future of Fiscal Policy:
American Economic Policy Debates in the 21\textsuperscript{st} Century
Place Based Policy

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Week 5

Thanks to Pat Kline and Juan Carlos Suárez Serrato for sharing notes/slides, much of which are reproduced here. Francesco Ruggieri provided excellent assistance making these slides.
Outline

1 Motivation

2 Policy
   - State and Local Business Tax Incentives

3 Theory
   - Rosen Roback Model
   - Kline Moretti Model
   - Place-based redistribution (Gaubert Kline Yagan (2019))

4 Evidence from Specific Place Based Policies
   - Empowerment Zones
   - Local Government Spending
   - Moving to Opportunity
   - Moved to Opportunity: Evidence from Public Housing Demolitions
   - Million Dollar Plants
   - Big Push: Tennessee Valley Authority
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Many programs target resources towards disadvantaged neighborhoods or regions.

In US, fed gov spends approx $15 B per year on spatial programs while state and local govts spend approx $80 B per year.

Glaeser and Gottleib (2008, BPEA):
- “The rationale for spending federal dollars to try to encourage less advantaged people to stay in economically weak places is itself extremely weak”

What is the economic case (if any) for targeting places instead of people?
California Enterprise Zone Program:
- “To stimulate economic development by providing tax incentives to businesses enabling private sector market forces to revive the local economy”

Empowerment Zones:
- “To create business opportunities and jobs in the most economically distressed areas of inner cities and the rural heartland”

Tennessee Valley Authority:
- “Touching and giving life to all forms of human concerns”
Motivation: Geographically concentrated economic activity

Figure 1  *Spatial distribution of economic output in the US, by square mile.* Notes: This figure reports the value of output produced in the US by square mile.

Source: Moretti (2011)
Motivation: Geographically concentrated poverty
Motivation: Geographically concentrated poverty/race

Motivation: Geographically concentrated poverty/race

The same data, aggregated by community area and shown with solid colors.

Motivation: Geographically concentrated shocks

The Parts of America Most Vulnerable to China

Some areas of the U.S. were hit especially hard by China’s rise, partly because those areas had lots of jobs in industries where imports surged the most.

Most-affected areas of the U.S.
Colors show which areas were most affected by China’s rise, based on the increase in Chinese imports per worker in each area from 1990 to 2007. Hovering over each area on the map will show a demographic breakdown of that area, below, and its most-affected industries, at right.

Most-affected industries
Most-affected industries, based on number of areas* Impact per worker†
Furniture and fixtures
Games, toys, and children’s vehicles
Sporting and athletic goods
Electronic components
Plastics products
Motor-vehicle parts and accessories
Electronic computers

* Most-affected industries, based on number of areas
† Impact per worker
Motivation: Geographically concentrated shocks

Furniture and fixtures

**Most-affected areas of the U.S.**

Colors show which areas were most affected by China’s rise, based on the increase in Chinese imports per worker in each area from 1990 to 2007. Hovering over each area on the map will show a demographic breakdown of that area, below, and its most-affected industries, at right.

<table>
<thead>
<tr>
<th>Most-affected 20%</th>
<th>Second-highest 20%</th>
<th>Middle 20%</th>
<th>Second-lowest 20%</th>
<th>Least-affected 20%</th>
</tr>
</thead>
</table>

**Most-affected industries**

<table>
<thead>
<tr>
<th>Most-affected industries</th>
<th>Impact per worker†</th>
</tr>
</thead>
<tbody>
<tr>
<td>Furniture and fixtures</td>
<td>196 areas $44k</td>
</tr>
<tr>
<td>Games, toys, and children’s vehicles</td>
<td>114 areas $488k</td>
</tr>
<tr>
<td>Sporting and athletic goods</td>
<td>106 areas $82k</td>
</tr>
<tr>
<td>Electronic components</td>
<td>87 areas $65k</td>
</tr>
<tr>
<td>Plastics products</td>
<td>84 areas $11k</td>
</tr>
<tr>
<td>Motor-vehicle parts and accessories</td>
<td>79 areas $12k</td>
</tr>
<tr>
<td>Electronic computers</td>
<td></td>
</tr>
</tbody>
</table>

Source: Autor Dorn Hanson [http://chinashock.info](http://chinashock.info)
Motivation: Geographically concentrated shocks
Motor-vehicle parts and accessories

Source: Autor Dorn Hanson http://chinashock.info
Motivation: Geographically concentrated recessions

Source: Yagan (2016)
Motivation: Geographically concentrated policy responses

Source: CBPP (2012)
Motivation: Geographically concentrated unemployment

<table>
<thead>
<tr>
<th>Rank</th>
<th>Metropolitan Area</th>
<th>Unemployment Rate (1)</th>
<th>Adjusted Unemployment Rate (2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Flint, MI</td>
<td>.1462</td>
<td>.1399</td>
</tr>
<tr>
<td>2</td>
<td>Yuba City, CA</td>
<td>.1099</td>
<td>.1072</td>
</tr>
<tr>
<td>3</td>
<td>Anniston, AL</td>
<td>.1074</td>
<td>.0899</td>
</tr>
<tr>
<td>4</td>
<td>Merced, CA</td>
<td>.1060</td>
<td>.0948</td>
</tr>
<tr>
<td>5</td>
<td>Toledo, OH/MI</td>
<td>.1058</td>
<td>.1064</td>
</tr>
<tr>
<td>6</td>
<td>Yakima, WA</td>
<td>.1047</td>
<td>.0970</td>
</tr>
<tr>
<td>7</td>
<td>Detroit, MI</td>
<td>.1044</td>
<td>.1082</td>
</tr>
<tr>
<td>8</td>
<td>Chico, CA</td>
<td>.1031</td>
<td>.1092</td>
</tr>
<tr>
<td>9</td>
<td>Modesto, CA</td>
<td>.1027</td>
<td>.1021</td>
</tr>
<tr>
<td>10</td>
<td>Waterbury, CT</td>
<td>.1023</td>
<td>.0918</td>
</tr>
</tbody>
</table>

Areas with the Lowest Rate

<table>
<thead>
<tr>
<th>Rank</th>
<th>Metropolitan Area</th>
<th>Unemployment Rate</th>
<th>Adjusted Unemployment Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>276</td>
<td>Provo-Orem, UT</td>
<td>.0391</td>
<td>.0369</td>
</tr>
<tr>
<td>277</td>
<td>Madison, WI</td>
<td>.0389</td>
<td>.0511</td>
</tr>
<tr>
<td>278</td>
<td>Odessa, TX</td>
<td>.0383</td>
<td>.0307</td>
</tr>
<tr>
<td>279</td>
<td>Fargo-Moorhead, ND/MN</td>
<td>.0362</td>
<td>.0467</td>
</tr>
<tr>
<td>280</td>
<td>Charlottesville, VA</td>
<td>.0348</td>
<td>.0362</td>
</tr>
<tr>
<td>281</td>
<td>Haver, Thibodaux LA</td>
<td>.0327</td>
<td>.0107</td>
</tr>
</tbody>
</table>
Motivation: Geographically concentrated unemployment

Differences are persistent ($\rho = .59$)

Source: Kline Moretti (2013)
Motivation: Geographically concentrated unemployment

Convergence is slowing

Source: Ganong and Shoag (2014)
Motivation: Geographically concentrated upward mobility

A. Absolute Upward Mobility: Average Child Rank for Below-Median Parents ($\bar{y}_{25}$) by CZ

Source: Chetty-Hendren-Kline-Saez (2014)
Effects on political polarization (and many other outcomes)

Source: Autor Dorn Hanson Majlesi (2017) http://chinashock.info. "Congressional districts exposed to larger increases in import penetration disproportionately removed moderate representatives from office in the 2000s. Trade-exposed districts with an initial majority white population or initially in Republican hands became substantially more likely to elect a conservative Republican, while trade-exposed districts with an initial majority-minority population or initially in Democratic hands became more likely to elect a liberal Democrat"
Stakes are high...

For low-income people, life expectancy is highest in California, New York, and Vermont. It is lowest in Nevada. The next 8 states with the lowest life expectancies form a belt connecting Michigan, Ohio, Indiana, Kentucky, Tennessee, Arkansas, Oklahoma, and Kansas.

Source: https://healthinequality.org
Stakes are high...

Life expectancy varies substantially across cities, especially for low-income people. For the poorest Americans, life expectancies are 6 years higher in New York than in Detroit. For the richest Americans, the difference is less than 1 year.
Summers case for place based policy

1. Widely uneven incidence of distress in US in terms of employment, opportunity, health, etc
2. National economic forces are doing little to cause this problem to solve itself
3. The propensity for migration to take place has diminished quite substantially and outmigrants are likely to be most able, skilled, catalytic
4. Disaffection of non-cosmopolitans who live away from major prosperous cities is a key source of protectionism, nationalism, anti-globalization, etc

Source: comments at October 2019 Boston Federal Reserve Conference
“I do not see how one can look at figures like these without seeing them as representing possibilities.” – Robert Lucas (1988)
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Poor states receive a lot of federal aid
Per capita spending and revenues vs per capita income

Source: Paul Krugman (2019)
Poor states receive a lot of federal aid

Source: Paul Krugman (2019)
Poor states receive a lot of federal aid

In Kentucky, for example, govt financed healthcare much bigger than coal employment

Source: Paul Krugman (2019)
Mean Earnings Disparities Across Place

A. Mean Earnings Disparities across Place

Source: Gaubert Kline Yagan (2019)
Mean Federal Income Tax Rate Disparities Across Place

B. Mean Federal Income Tax Rate Disparities across Place

Source: Gaubert Kline Yagan (2019)
State and Local economic development spending
A substantial portion is for local business incentives. Source Barik (2019)

<table>
<thead>
<tr>
<th>CURRENT PROGRAMS</th>
<th>Annual dollars (in billions)</th>
</tr>
</thead>
<tbody>
<tr>
<td>State and local business tax incentives and other cash incentives</td>
<td>46.3</td>
</tr>
<tr>
<td>Customized training programs</td>
<td>0.6</td>
</tr>
<tr>
<td>Other state economic development programs</td>
<td>2.8</td>
</tr>
<tr>
<td><strong>Subtotal, state/local programs</strong></td>
<td><strong>49.7</strong></td>
</tr>
<tr>
<td>Manufacturing extension (federal/state/fees)</td>
<td>0.4</td>
</tr>
<tr>
<td>Economic Development Administration (EDA)</td>
<td>0.3</td>
</tr>
<tr>
<td>Economic development portion of HUD’s Community Development Block Grants</td>
<td>1.1</td>
</tr>
<tr>
<td>Small Business Administration</td>
<td>0.8</td>
</tr>
<tr>
<td>Other economic development programs in USDA, HUD, Commerce</td>
<td>2.0</td>
</tr>
<tr>
<td><strong>Subtotal, mostly federal spending</strong></td>
<td><strong>4.6</strong></td>
</tr>
<tr>
<td>Opportunity Zones tax credits</td>
<td>1.5</td>
</tr>
<tr>
<td>New markets tax credit</td>
<td>1.4</td>
</tr>
<tr>
<td>Other tax expenditures that might promote local economic development</td>
<td>2.3</td>
</tr>
<tr>
<td><strong>Subtotal, federal tax expenditures</strong></td>
<td><strong>5.3</strong></td>
</tr>
<tr>
<td><strong>Total of federal programs and tax expenditures</strong></td>
<td><strong>9.9</strong></td>
</tr>
<tr>
<td><strong>Total of all levels of government</strong></td>
<td><strong>59.6</strong></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>PAST PROGRAMS</th>
</tr>
</thead>
<tbody>
<tr>
<td>Empowerment zones/enterprise communities (peak annual activity in early 2000s)</td>
</tr>
<tr>
<td>Appalachian Regional Commission (peak annual spending 1966–1975)</td>
</tr>
<tr>
<td>Tennessee Valley Authority (peak annual spending 1950–1955)</td>
</tr>
</tbody>
</table>
State and Local Business Tax Incentives
Evaluating State and Local Business Tax Incentives

- State and local governments spend billions of dollars each year on tax incentives and subsidies to attract and retain firms (Bartik, 2017, Slattery, 2019)

- Incentive policies are highly controversial
  - Attracting industrial activity is key for local economic growth and prosperity
  - Others question incentive spending effectiveness and mounting costs

- Evaluating these incentives requires overcoming three challenges
  1. Data limitations: difficult to measure prevalence, size, and composition of incentives
  2. Lack of transparency: hard to determine selection process
  3. Do not observe how economic activity would have evolved in the absence of deals
Objective of state and local government is to maximize local welfare

\[ V = \sum_{i \in \text{workers}} \psi_i^W V_i^W + \sum_{i \in \text{owners}} \psi_i^O V_i^O + \sum_{i \in \text{politicians}} \psi_i^P V_i^P \]  

- **Workers**: \( V_i^W = w - p - t + g \) is real wages less taxes plus gov amenities
  - Higher wages, higher local prices, higher taxes, less \( g \). Big wage gain if unemployed.
  - \( \psi_i \) is individual \( i \)'s social welfare weight
- **Owners**: \( V_i^O = (1 - t_{\text{corp}} + \text{incentive}_i) \text{profits}_i \)
  - Higher factor costs, higher product demand, higher taxes, less \( g \).
- **Politicians**
  - Re-election odds, campaign contributions, pork, etc

**Policy Instruments**

- Firm-specific tax incentive
- Lower state corporate tax rate
- Narrow state corporate tax base (e.g., provide a state investment tax credit)
- Many others
Welfare Considerations

1. Marginal Value of Public Funds Approach for each policy instrument
   - WTP for the policy (incorporating social welfare weight of that group)
   - Fiscal externalities

2. Direct approach
   - \( \frac{dV}{dPolicy} \)
   - Effects on different groups
   - Effects on factor prices (boost for labor and land), output prices (negative congestion effects)
   - Increase in net tax burden on local residents and/or lower government services
   - Deadweight loss from higher local tax burden
Benefits and Costs of Different Instruments

1. Firm-specific subsidies
   - Some places might get more location-specific value
   - Can better target mobile firms (or not extract rents from new firms) or firms with more spillovers
   - Political economy benefits: more certainty, pork, salience

2. But
   - Hard to know which firms are inframarginal (the “but for” debate)
   - Hard to “pick winners”, allocation of spending may be more about politics than economics
   - Lower tax revenue and lower public goods
   - Congestion, higher factor prices, etc
Outline

1 Policy: what are firm-specific subsidies?
   - Example 2008 Volkswagen Deal
   - Comprehensive data on firm-specific subsidies
   - Typical deal characteristics

2 Selection
   - Which firms receive them?
   - Which places provide firm-specific subsidies?
   - Why do they provide firm-specific subsidies?

3 Impacts: what are the effects on employment and economic activity?
   - Approach: compare outcomes in winner versus runner up locations
   - Event study of 2008 Volkswagen Deal
   - Event study of all firm-specific subsidies
2008 Volkswagen Deal in Tennessee

“There’s nothing quite like the automobile industry to bring in money, raise family incomes and bring in jobs”

- VW build new assembly plant in Chattanooga with 2,000 emps and $1B spending
- TN grants VW a subsidy worth **$558 million**
  - Local property tax abatements over 30 years ($200M)
  - Enhanced state job and investment tax credits over 20 years ($200M)
  - Property given to VW ($81M)
  - Worker training ($30M)
  - Highway and road construction ($43M) + Rail line upgrades ($3.5M)
- TN promises specialized tax credits for any neighboring suppliers
- Location decision was “truly a very close competition”: the runner-up was a site annexed by Huntsville, AL, where the subsidy offer was at least **$386 million**
- TN projected VW would have $100M in annual payroll, help create 14,000 total jobs, and have a total economic benefit of $600M per year
## Average Deal and Top Industries Receiving Subsidies

<table>
<thead>
<tr>
<th>Industry (NAICS)</th>
<th>Subsidy ($ M)</th>
<th># Jobs Promised</th>
<th>Cost per Job ($)</th>
<th># of Deals</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>Median</td>
<td>Mean</td>
<td>Median</td>
</tr>
<tr>
<td>Full sample</td>
<td>153.0</td>
<td>57.2</td>
<td>1.888</td>
<td>850</td>
</tr>
<tr>
<td>Analysis sample</td>
<td>166.7</td>
<td>62.8</td>
<td>1.495</td>
<td>900</td>
</tr>
<tr>
<td>Manufacturing analysis sample</td>
<td>223.6</td>
<td>77.9</td>
<td>1.435</td>
<td>850</td>
</tr>
<tr>
<td>Automobile manuf. (3361)</td>
<td>302.9</td>
<td>140.9</td>
<td>4.850</td>
<td>2,000</td>
</tr>
<tr>
<td>Aerospace manuf. (3364)</td>
<td>338.5</td>
<td>94.9</td>
<td>4.989</td>
<td>1,100</td>
</tr>
<tr>
<td>Semiconductor/electronic manuf. (3344)</td>
<td>281.1</td>
<td>97.1</td>
<td>672</td>
<td>500</td>
</tr>
<tr>
<td>Financial activities (5239)</td>
<td>96.1</td>
<td>24.9</td>
<td>2,582</td>
<td>1,691</td>
</tr>
<tr>
<td>Pharmaceutical/medicine manuf. (3254)</td>
<td>61.3</td>
<td>46.7</td>
<td>601</td>
<td>500</td>
</tr>
<tr>
<td>Scientific R&amp;D svc (5417)</td>
<td>122.7</td>
<td>61.1</td>
<td>568</td>
<td>548</td>
</tr>
<tr>
<td>Basic chemical manuf. (3251)</td>
<td>315.7</td>
<td>48.5</td>
<td>187</td>
<td>130</td>
</tr>
<tr>
<td>Rubber product manuf. (3262)</td>
<td>120.0</td>
<td>97.8</td>
<td>1,472</td>
<td>1,450</td>
</tr>
<tr>
<td>Information Technology (5415)</td>
<td>162.6</td>
<td>33.7</td>
<td>2,548</td>
<td>800</td>
</tr>
<tr>
<td>Data processing, hosting/related svc (5182)</td>
<td>159.9</td>
<td>106.9</td>
<td>463</td>
<td>100</td>
</tr>
<tr>
<td>Petroleum/coal manuf. (3241)</td>
<td>131.3</td>
<td>84.3</td>
<td>1,389</td>
<td>300</td>
</tr>
</tbody>
</table>

Notes: Tabulated using firm-level subsidy data from Slattery (2019). The full sample is 535 subsidy deals over the period of 2002-2017. The subsidy size is normalized to 2017 USD and a 10 year contract. The eleven top industries listed above make up 46% of the sample in terms of number of deals, and 65.5% of the sample in terms of dollars spent.

Source: Slattery Zidar (2019)
Firms Receiving Discretionary Subsidies Are Larger

<table>
<thead>
<tr>
<th>Jobs promised</th>
<th>All Discretionary Subsidies (2002-2016)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td># Subsidies</td>
</tr>
<tr>
<td>1 - 99</td>
<td>36</td>
</tr>
<tr>
<td>100 - 249</td>
<td>44</td>
</tr>
<tr>
<td>250 - 499</td>
<td>77</td>
</tr>
<tr>
<td>500 - 999</td>
<td>140</td>
</tr>
<tr>
<td>1000+</td>
<td>238</td>
</tr>
</tbody>
</table>

- Most establishments that enter are small (1-99 employess)
- Firms are more likely to receive a discretionary subsidy when they build a new establishment with 1000+ employees

Source: Slattery Zidar (2019)
They also have Greater Employment, Profits, Revenue and Capital Stock

<table>
<thead>
<tr>
<th></th>
<th>All Compustat</th>
<th>Subsidized Firms</th>
<th>Subsidized Firms: Year of Deal</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>Median</td>
<td>Mean</td>
</tr>
<tr>
<td>Employees (1000s)</td>
<td>9.0</td>
<td>0.6</td>
<td>72.0</td>
</tr>
<tr>
<td>Capital Stock ($M)</td>
<td>1,514.4</td>
<td>28.2</td>
<td>12,098.3</td>
</tr>
<tr>
<td>Revenue ($M)</td>
<td>3,461.8</td>
<td>184.5</td>
<td>40,095.5</td>
</tr>
<tr>
<td>Gross Profit ($M)</td>
<td>1,139.8</td>
<td>67.5</td>
<td>13,239.3</td>
</tr>
<tr>
<td>Market Value ($M)</td>
<td>2,997.1</td>
<td>189.5</td>
<td>45,988.1</td>
</tr>
<tr>
<td>State Income Taxes ($M)</td>
<td>5.1</td>
<td>0.0</td>
<td>57.8</td>
</tr>
<tr>
<td>Total Income Taxes ($M)</td>
<td>99.7</td>
<td>1.0</td>
<td>1,239.1</td>
</tr>
<tr>
<td>Observations</td>
<td>107,219</td>
<td>2,470</td>
<td>313</td>
</tr>
</tbody>
</table>

Notes: This table includes descriptive statistics on all firms included in Compustat, 2002-2014, and the Compustat firms that received discretionary subsidies. 61% of the firms receiving discretionary subsidies were found in Compustat. Dollars are measured in 2016 dollars.

Source: Slattery Zidar (2019)
### Which Places Provide Firm-Specific Subsidies?

**Unique counties in 2000**

<table>
<thead>
<tr>
<th>County:</th>
<th>Winner (Full)</th>
<th>Winner (Analysis)</th>
<th>Runner-up</th>
<th>Average</th>
<th>Pop &gt; 100K</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>Median</td>
<td>Mean</td>
<td>Median</td>
<td>Mean</td>
</tr>
<tr>
<td>Employment (K)</td>
<td>940.9</td>
<td>452.0</td>
<td>481.3</td>
<td>442.6</td>
<td>1,233.4</td>
</tr>
<tr>
<td>Population (K)</td>
<td>1,955.2</td>
<td>884.4</td>
<td>959.4</td>
<td>860.8</td>
<td>2,566.8</td>
</tr>
<tr>
<td>Average wages (K)</td>
<td>54.0</td>
<td>51.2</td>
<td>55.3</td>
<td>50.7</td>
<td>57.2</td>
</tr>
<tr>
<td>Population density</td>
<td>3,051.8</td>
<td>1,306.1</td>
<td>3,560.3</td>
<td>1,213.9</td>
<td>3,555.3</td>
</tr>
<tr>
<td>% emp in mfg.</td>
<td>15.0</td>
<td>14.5</td>
<td>14.0</td>
<td>13.8</td>
<td>14.1</td>
</tr>
<tr>
<td>% emp info &amp; prof svcs.</td>
<td>27.6</td>
<td>28.6</td>
<td>29.3</td>
<td>29.2</td>
<td>30.1</td>
</tr>
<tr>
<td>% urban</td>
<td>92.4</td>
<td>97.1</td>
<td>93.5</td>
<td>96.6</td>
<td>95.0</td>
</tr>
<tr>
<td>% Bachelor’s or more</td>
<td>27.6</td>
<td>25.9</td>
<td>29.6</td>
<td>27.0</td>
<td>29.6</td>
</tr>
<tr>
<td>% white</td>
<td>69.3</td>
<td>70.5</td>
<td>72.9</td>
<td>73.1</td>
<td>67.0</td>
</tr>
<tr>
<td>% Hispanic</td>
<td>16.7</td>
<td>10.5</td>
<td>13.8</td>
<td>8.2</td>
<td>18.2</td>
</tr>
<tr>
<td>% foreign-born</td>
<td>14.9</td>
<td>11.2</td>
<td>13.2</td>
<td>10.9</td>
<td>17.1</td>
</tr>
<tr>
<td>log housing units</td>
<td>12.8</td>
<td>12.8</td>
<td>12.6</td>
<td>12.8</td>
<td>13.2</td>
</tr>
<tr>
<td>log median house value</td>
<td>12.1</td>
<td>12.0</td>
<td>12.1</td>
<td>12.0</td>
<td>12.2</td>
</tr>
<tr>
<td>Wage bill (M)</td>
<td>54,293.2</td>
<td>24,556.1</td>
<td>28,296.1</td>
<td>22,566.0</td>
<td>72,235.9</td>
</tr>
<tr>
<td>Personal income (M)</td>
<td>92,498.7</td>
<td>41,676.7</td>
<td>50,526.2</td>
<td>41,523.0</td>
<td>123,204.7</td>
</tr>
<tr>
<td>Personal inc/capita (K)</td>
<td>48.3</td>
<td>43.8</td>
<td>50.9</td>
<td>46.0</td>
<td>51.0</td>
</tr>
<tr>
<td>Unemployment rate (%)</td>
<td>3.9</td>
<td>3.7</td>
<td>3.5</td>
<td>3.5</td>
<td>3.9</td>
</tr>
</tbody>
</table>

**Observations** 268 115 125 496 494

*Notes: In the full sample, 268 unique counties win the 535 deals. In the analysis sample, 115 win 195 deals.*

**Source:** Slattery Zidar (2019)
Firm-Specific Subsidies are Larger and More Generous in Lower Wage Locations

Subsidies Relative to Average Wages

Subsidy per Job Relative to Average Wages

Notes: Average wages are measured in the year of the firm-specific deal. Triangles in plot are individual data points; circles are binned data. Best fit line estimates are taken from population-weighted linear regression of y-axis variable onto x-axis variable.

Source: Slattery Zidar (2019)
Determinants of Subsidy Size: Evidence from North Carolina

- Subsidy ↑ $29K per direct job
- Subsidy is higher for high-wage firms
- Subsidy ↑ $100 for $1 of average wages → going from wage $64K (mean) to $125K (P95) gives ≈ 2X subsidy
- NC predicts a multiplier of 1.4 for a median subsidy deal and 2.5 for every new skilled tradable job (close to local multiplier effects in Moretti (2010))

Source: Slattery Zidar (2019)
North Carolina’s projected effects of firm subsidies on tax revenue and GDP

- NC projects recovering $0.76 of tax revenue and $68 of GDP per dollar of subsidy

Source: Slattery Zidar (2019)
2008 Volkswagen Deal: winner vs runner up

Impact of the 2008 Volkswagen deal on local auto employment (NAICS 336)

Employment in Transportation Equipment Manufacturing

Differences in Employment Between Winner and Runner-up

Source: Slattery Zidar (2019)
Event Study: Impact of Winning a Firm-Specific Deal on 3-D Employment
Local Employment in 3-Digit Industry of Deal

Logs

Levels

Notes: The event studies control for log population, log employment and log average wages 10 years before the deal, and for deal year fixed effects. Standard errors clustered at the state level.

Source: Slattery Zidar (2019)
Spillovers: Local Employment in 3-D, 2-D and 1-D Industry of Deal

Logs

Levels

Notes: The event studies control for log population, log employment and log average wages 10 years before the deal, and for deal year fixed effects. Standard errors clustered at the state level.

Source: Slattery Zidar (2019)
Welfare effects of subsidy deals are hard to measure

1. Some inputs to welfare analysis are (now) easier to measure
   - Mechanical Costs: average discretionary subsidy of $153M is $15.3M/year

2. Harder to measure:
   - Firm Location: inframarginal vs marginal firms (e.g., Amazon in NYC)?
   - Externalities: hard to detect indirect jobs, or IO linkages, price effects, congestion
   - Fiscal Externalities: hard to measure both state and local budget effects
     - 1000 jobs $\times$ 65K $\approx$ 65M wagebill. At 5-10% tax rates, rev of $\approx$3-7M
     - But unlikey there are 1000 net new jobs. Most do not go from unemployed to 65K.

3. How are these financed? Marginal value of those taxes and/or reduced spending?
   - Hard to find systematic data, but some case studies suggest the following sources:
     - Foregone future revenues (e.g., Foxconn)
     - Reduce other government spending and/or raise taxes (e.g., VA, TX, FL)
General considerations:

- Production efficiency/misallocation vs price discrimination/second best?
  - Distorting the allocation of resources across firms and locations?
  - Taxing more elastic firms less than the full corporate rate (Ramsey rule)?
  - Offsetting existing distortions / effectively subsidizing poor places?

- Beggar-thy-neighbor vs match effects?
  - Zero sum unless really big match effects?

- Importance of political vs economic considerations?
Outline

1. Motivation

2. Policy
   - State and Local Business Tax Incentives

3. Theory
   - Rosen Roback Model
   - Kline Moretti Model
   - Place-based redistribution (Gaubert Kline Yagan (2019))

4. Evidence from Specific Place Based Policies
   - Empowerment Zones
   - Local Government Spending
   - Moving to Opportunity
   - Moved to Opportunity: Evidence from Public Housing Demolitions
   - Million Dollar Plants
   - Big Push: Tennessee Valley Authority
Rosen Roback Model
Overview

Goals
- Characterize effect of amenity s change on prices (wages and rents)
- Infer the value of amenities

Markets
- Labor: price $w$, quantity $N$
- Land: price $r$, quantity $L = L^w + L^p$ for workers and production
- Goods: price $p = 1$, quantity $X$

Agents
- Workers (homogenous, perfectly mobile)
- Firm (perfectly competitive, CRS)

Indifference Conditions
- Workers have same indirect utility in all locations
- Firm has zero profit (i.e., unit costs equal 1)
Aside: Components of Models

Three parts of any model

1. Exogenous parameters: model elements that are taken “as given”
2. Endogenous outcomes: model elements that “move around”
3. Equilibrium conditions: the set of rules that tells you what the endogenous model outcomes should be for a given set of exogenous model parameters.

“Given a [insert set of exogenous model parameters here], equilibrium is defined by the [insert endogenous model outcomes here] such that [list equilibrium conditions here].”
Exogenous parameters

- **Workers Parameters:** $s, \theta_W, \gamma, I$
  - $s$ is level of amenities
  - $\theta_W$ governs importance of amenities for utility
  - $\gamma$ governs importance of goods for utility
  - $1 - \gamma$ governs importance of land for utility
  - $I$ is non-labor income

- **Firm Parameters:** $s, \theta_F, \alpha$
  - $s$ is level of amenities
  - $\theta_F$ governs importance of amenities for productivity
  - $\alpha$ is output elasticity of labor
  - $1 - \alpha$ is output elasticity of land
Endogenous Model Outcomes

- Labor: price $w$, quantity $N$
- Land: price $r$, quantities $L^w, L^p$ for workers and production
- Goods: price $p = 1$, quantity $X$

so endogenous outcomes are $w, r, N, L^w, L^p, X$
Equilibrium Concept: Two key indifference conditions

In equilibrium, workers and firms are indifferent across cities with different levels of \( s \) and endogenously varying wages \( w(s) \) and rents \( r(s) \):

\[
\begin{align*}
c(w(s), r(s), s) &= 1 \\
V(w(s), r(s), s) &= V^0
\end{align*}
\]

where \( V^0 \) is the initial equilibrium level of indirect utility.

Specifically, in our example:

*Given* \( s, \theta_W, \theta_F, \gamma, I, \alpha \), *equilibrium is defined by local prices and quantities* \( \{w, r, N, L^w, L^p, X\} \) *such that 2 and 3 hold and land markets clear.*

N.B. We will mainly be focusing on prices: \( w(s) \) and \( r(s) \).
Solving for effect of amenity changes on prices

- Differentiate 2 and 3 with respect to $s$ and rearrange, we have:

$$\begin{bmatrix} c_w & c_r \\ V_w & V_r \end{bmatrix} \begin{bmatrix} w'(s) \\ r'(s) \end{bmatrix} = \begin{bmatrix} -c_s \\ -V_s \end{bmatrix}$$

(4)

- Solving for $w'(s), r'(s)$, we have

$$w'(s) = \frac{V_r c_s - c_r V_s}{c_r V_w - c_w V_r}$$

$$r'(s) = \frac{V_s c_w - c_s V_w}{c_r V_w - c_w V_r}$$
Effect of amenity changes on prices

- Special cases of interest:
  1. Amenity only valued by consumers: $\theta_F = 0 \Rightarrow c_s = 0$
  2. Amenity only has productivity effect: $\theta_W = 0 \Rightarrow V_s = 0$
  3. Firms use no land $1 - \alpha = 0$ and amenity is non-productive $\theta_F = 0$: $c(w(s)) = 1$, $c_r = c_s = 0$
1. Amenity only valued by consumers: $\theta_F = 0 \Rightarrow c_s = 0$

- When $c_s = 0$, higher $s \Rightarrow$ higher $r$, lower $l$

- Workers are willing to pay more in land rents and receive less in pay to have access to higher levels of amenities
2. Amenity only has productivity effect: $\theta_W = 0 \Rightarrow V_s = 0$

- When $V_s = 0$, higher $s \Rightarrow$ higher $r$ and higher $l$
- Firms are willing to pay more in land rents and wages to access higher productivity due to amenities
3. Firms use no land $\alpha = 0$, amenity not productive $\theta_F = 0$

- $\frac{V_s}{V_w}$ = marginal WTP for a change in $s$ so the marginal value of a change in the amenity is “fully capitalized” in rents
Aside: evidence of the value of local public goods

![Graph showing mean log house prices vs. year relative to election]

**Figure VI**
Recursive, One-Step, and Forward-Looking Estimates of Dynamic TOT Effects of Bond Passage on Log House Prices, by Years since Election

Source: Cellini, Ferreira, Rothstein (2010)
Rationales for place-based policies

**Equity**

1. Economists have generally been skeptical of equity-based arguments, as location is being used to serve a person-based motive: subsidizing poor households (see Glaser and Gottlieb, 2008)
2. Could do so more directly through tax progressive or transfer programs
3. Mobility can undermine spatial targeting. Rosen-roback model (with mobile workers and inelastic housing supply) predicts that entire benefit of location-based subsidies will be capitalized into land rents
4. However, if workers (or firms) are less mobile, redistributive policies can benefit inframarginal workers (firms)

**Efficiency:** Can remedy market failures

1. Public Goods (amenities like public safety or productive public goods like roads)
2. Agglomeration
3. Labor market frictions
4. Missing insurance/credit markets
5. Pre-existing distortions
Kline Moretti Model
Overview

1. Goals
   - Characterize effect of place-based wage subsidy on prices (wages and rents), city size, and welfare
   - Determine aggregate benefits (costs) and how they are distributed across agents and locations

2. Two Locations $c \in \{a, b\}$

3. Markets
   - Local labor and housing: price $w_c$, quantity $N_c$. Price $r_c$, $N_c$
   - Global capital and goods: price $\rho$, quantity $K_c$. Price $p = 1$, $Y_c$

4. Agents
   - Workers (continuum, have heterogeneous taste draws)
   - Landlord (representative, housing has upward sloping supply)
   - Firm (perfectly competitive, CRS, traded good)
   - Government provides ad valorem wage credit $\tau_c$ to firms

5. Key Indifference Condition
   - *Marginal worker* has same indirect utility in both locations
Preferences

- Indirect utility of individual $i$ in location $c$ is given

$$U_{ic} = w_c - r_c + A_c - t + e_{ic} = v_c$$

- where
  - nominal wages $w_c$
  - cost of housing $r_c$
  - lump sum taxes $t$
  - local amenities $A_c$
  - common indirect utility component $v_c$
  - $e_{ic}$ represents worker $i$’s idiosyncratic preferences for location $c$

Source: Kline (2017)
Location Choice

Aside on Discrete Choice

• Suppose two cities: \( a \) and \( b \)
• Household chooses city \( a \) if and only if
  \[ e_{ib} - e_{ia} < v_{ib} - v_{ia} \]
• The fraction of workers locating in city \( a \) can be expressed as:
  \[ N_a = \Lambda \left( \frac{v_a - v_b}{s} \right) \]
• \( s \) is scale parameter
  • \( s \to 0 \) “skating-rink” model ala Roback (1982)
  • \( s \to \infty \) immobility
• Special case: \( \Lambda(\cdot) = \frac{\exp(\cdot)}{1+\exp(\cdot)} \) is the standard logistic CDF
• \( N_a \) is increasing in:
  • the real-wage gap: \((w_a - r_a) - (w_b - r_b)\)
  • the amenity gap: \(A_a - A_b\)
Housing Supply

- Absentee landlords
- Constant elasticity inverse housing supply function:
  \[ r_c = z_c N_c^{k_c} \]
  - \( z_c \) governs housing productivity
  - \( k_c \) governs the elasticity of housing supply
- Landlord profits are constant fraction of total rents
  \[ \Pi_c = \frac{k_c}{k_c + 1} r_c N_c \]

Source: Kline (2017)
Production and price taking

- Firms produce a single good $Y$ using labor and a local amenity
- $Y$ is a traded good sold on international markets at price $1$
- Cobb-Douglas production function with constant returns to scale:
  \[ Y_c = X_c N_c^\alpha K_c^{1-\alpha} \]
  where:
  - $X_c$ is a city-specific productivity shifter
  - $N_c$ is the fraction of workers in community $c$
  - $K_c$ is the local capital stock
- Firms can rent as much capital as desired at fixed price $\rho$

Source: Kline (2017)
A place based policy

- The government provides an ad valorem wage credit $\tau_c$ to employers in community $c$
- Financed by lump sum taxes on all workers in both locations
- Balanced budget constraint:
  \[ w_a \tau_a N_a + w_b \tau_b N_b = t \]
- Firms equate the marginal revenue product of labor to wages net of taxes:
  \[ w_c (1 - \tau_c) = \alpha \frac{Y_c}{N_c} \]
- First-order condition for capital:
  \[ \rho = (1 - \alpha) \frac{Y_c}{K_c} \]

Source: Kline (2017)
Labor demand

- Inverse labor demand schedule in location $c$:

$$\ln w_c = C + \frac{\ln X_c}{\alpha} - \frac{1 - \alpha}{\alpha} \ln \rho - \ln(1 - \tau_c)$$

where $C \equiv \ln \alpha + \frac{1 - \alpha}{\alpha} \ln(1 - \alpha)$

- Inverse labor demand is *horizontal* in wage-employment space due to:
  - production function with constant returns to scale
  - elastic supply of capital at price $\rho$
  - firms make zero profits

Source: Kline (2017)
Labor labor market equilibrium

- Equilibrium: the marginal worker’s relative preference for city $b$ over city $a$ equals the difference in real wages net of amenities:

\[
\text{Taste Differences} = \frac{e^C}{\rho^{\frac{1}{1-\alpha}}} \left( \frac{X_a^{\alpha}}{1-\tau_a} - \frac{X_b^{\alpha}}{1-\tau_b} \right) + \\
\text{Wage difference} - \left( z_a N_a^k - z_b (1 - N_a)^k \right) + \frac{A_a - A_b}{\text{Amenity difference}}
\]

- LHS: quantiles of workers’ relative preferences $(e_{ib} - e_{ia})$ for city $b$ as a function of $N_a$ ⇒ supply curve to city $a$

- RHS: difference in mean utilities between the two communities ⇒ relative demand curve for residence in city $a$ vs. city $b$

Source: Kline (2017)
A subsidy raises city size

Source: Kline (2017)
but decreases welfare (especially for workers)

Source: Kline (2017)
Who wins?

Source: Kline (2017)

Future of Fiscal Policy (Econ 593i)

Week 5
Efficiency costs

- Average worker utility

\[ V = E \max\{U_{ia}, U_{ib}\} = s \log \left( \exp \left( \frac{v_a}{s} \right) + \exp \left( \frac{v_b}{s} \right) \right) \]

- For small subsidy, impact on welfare is impact on after-tax disposable income

\[ \frac{dV}{d\tau_a} = N_a \frac{d(w_a - r_a)}{d\tau_a} + N_b \frac{d(w_b - r_b)}{d\tau_a} - \frac{dt}{d\tau_a} \]

- Net impact on worker utility + landlord profits

\[ \frac{d(V + \Pi_a + \Pi_b)}{d\tau_a} = -\eta N_a \tau_a \]

where \( \eta = -\frac{dN_a}{d(1-\tau_a)} \frac{1-\tau_a}{N_a} \geq 0 \) gives mobility elasticity

- Harberger (1964) “triangle” approximation: \( DWL \approx \frac{1}{2} \eta \tau_a^2 N_a \)

Source: Kline (2017)
Subsidizing a place yields a transfer to targeted households (and landlords) but distorts location decisions.

Efficient transfer: no quantity response / job creation!

Ramsey (1927)-style targeting principle: subsidize locations that are least elastic.

Empirical question: when are elasticities big?

Source: Kline (2017)
Place-based redistribution (Gaubert Kline Yagan)
Can PBR deliver the same equity benefit as the income tax at lower efficiency cost?

PBR is likely to raise welfare under four intuitive conditions:

1. PBR has small efficiency costs when the mobility response to place-based transfers is small.
2. PBR has small efficiency costs when productivity differs little between Distressed and Elsewhere for workers who move in response to tax changes.
3. The income tax has large efficiency costs when labor supply is especially elastic.
4. PBR is likely to be desirable when high-earners sort disproportionately into Elsewhere and low-earners into Distressed because under these conditions the place-based transfer will be especially well targeted: it redistributes from rich to poor without the income tax’s efficiency costs among non-movers.

In sum, PBR, like income taxation, generates efficiency costs to be traded off against equity gains. Because PBR and income-based taxation trigger qualitatively different behavioral responses, PBR will be a useful complement to income taxation in some empirical settings - while it could be superfluous in others.
Outline

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   - Local Government Spending
   - Moving to Opportunity
   - Moved to Opportunity: Evidence from Public Housing Demolitions
   - Million Dollar Plants
   - Big Push: Tennessee Valley Authority
Empowerment Zones
Assessing a Prominent Place Based Policy (Busso et al.)

Case Study: Empowerment Zones

[Diagrams of Detroit and Chicago with Empowerment Zones highlighted]
What is the incidence of Round I of the federal urban Empowerment Zone (EZ) program?

Evidence helps determine whether or not place based policies are effective in accomplishing their goals.

BGK conduct the first microfounded equilibrium welfare evaluation of a large-scale place based policy.
The EZ program is a series of incentives to encourage investment in the neediest urban and rural areas.

It consists of spatially targeted investments, such as employment tax credits and block grants.

### Table 1—1990 Characteristics of First Round Empowerment Zones (EZ)

<table>
<thead>
<tr>
<th>City</th>
<th>Total population</th>
<th>Population rank</th>
<th>Population in EZ</th>
<th>Poverty rate in EZ</th>
<th>Unemployment rate in EZ</th>
<th>EZ area (square miles)</th>
<th>Number of census tracts</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atlanta</td>
<td>395,337</td>
<td>37</td>
<td>43,792</td>
<td>58</td>
<td>20</td>
<td>8.1</td>
<td>20</td>
</tr>
<tr>
<td>Baltimore</td>
<td>736,014</td>
<td>13</td>
<td>72,725</td>
<td>42</td>
<td>16</td>
<td>7.1</td>
<td>23</td>
</tr>
<tr>
<td>Chicago</td>
<td>2,783,484</td>
<td>3</td>
<td>200,182</td>
<td>49</td>
<td>28</td>
<td>14.3</td>
<td>81</td>
</tr>
<tr>
<td>Detroit</td>
<td>1,027,974</td>
<td>7</td>
<td>106,273</td>
<td>47</td>
<td>28</td>
<td>19.5</td>
<td>42</td>
</tr>
<tr>
<td>New York</td>
<td>7,320,621</td>
<td>1</td>
<td>204,625</td>
<td>42</td>
<td>18</td>
<td>6.3</td>
<td>51</td>
</tr>
<tr>
<td>Philadelphia/</td>
<td>1,594,339</td>
<td>5</td>
<td>52,440</td>
<td>50</td>
<td>23</td>
<td>4.3</td>
<td>17</td>
</tr>
</tbody>
</table>

Source: 1990 Decennial Census and HUD.
Assessing a Prominent Place Based Policy (Busso et al.)

Program Benefits

- **Employment tax credit**
  - EZ employers were eligible for a credit of up to 20 percent of the first $15,000 in wages paid to each employee who lived *and worked* in the EZ.
  - Roughly 20% wage subsidy!

- **Social Services Block Grant Funds (SSBG)**
  - Each EZ became eligible for $100 million in SSBG funds.
  - Could be used for: infrastructure investment, improving access to credit, job training programs, childcare programs, promotion of homeownership, emergency housing assistance, etc.
Assessing a Prominent Place Based Policy (Busso et al.)

Methods: Empirical Strategy

- Empirical strategy involves comparing EZ neighborhoods to rejected and future zones using a difference-in-differences estimator

\[ \Delta Y_{tzc} = \beta T_z + X_{n(t)}' \alpha^x + P_c' \alpha^p + e_{tzc} \]

- \( \Delta Y_{tzc} \) is change in outcome in tract \( t \) of zone \( z \) in city \( c \)
- \( T_z \) is an indicator for EZ status
- \( P_c \) is a vector of city characteristics
- \( X_{n(t)} \) is a vector of proxies for trends in productivity and amenities
Data

- Household and establishment panel data comes from the Census, the Standard Statistical Establishment List (SSEL), and the Longitudinal Business Database (LBD)

- First-round EZ applications were obtained from the US Department of Housing and Urban Development

- Housing price data is from the Office of Federal Housing Enterprise Oversight (OFHEO)
### Table 2: Selected Effects of Round I Empowerment Zone Designations, 1990-2000

<table>
<thead>
<tr>
<th>Outcome</th>
<th>Estimated Effect</th>
</tr>
</thead>
<tbody>
<tr>
<td>Log of Jobs (data from Longitudinal Business Database)</td>
<td>0.179***</td>
</tr>
<tr>
<td>Log of Jobs (data from U.S. Census)</td>
<td>0.145*</td>
</tr>
<tr>
<td>Log of Zone Jobs Held by Zone Residents</td>
<td>0.150</td>
</tr>
<tr>
<td>Log of Zone Jobs Held by Nonresidents</td>
<td>0.097</td>
</tr>
<tr>
<td>Log of Weekly Wage Income of Zone Residents</td>
<td>0.053**</td>
</tr>
<tr>
<td>Log of Weekly Wage Income of Zone Workers</td>
<td>0.017</td>
</tr>
<tr>
<td>Log of Weekly Wage Income of Zone Residents Working in Zone</td>
<td>0.133**</td>
</tr>
<tr>
<td>Log of Weekly Wage Income of Nonresidents Working in Zone</td>
<td>0.005</td>
</tr>
<tr>
<td>Log of Rent</td>
<td>0.006</td>
</tr>
<tr>
<td>Log of House Value</td>
<td>0.281**</td>
</tr>
<tr>
<td>Log of Population</td>
<td>0.028</td>
</tr>
<tr>
<td>Percentage Black</td>
<td>-0.011</td>
</tr>
<tr>
<td>Percentage with College Degree*</td>
<td>0.020***</td>
</tr>
</tbody>
</table>

Notes: Estimated impacts derived from regression-adjusted difference-in-differences model. Statistical significance levels based on a Wild bootstrap t-test are indicated as *** 1 percent; ** 5 percent; * 10 percent. For more details, see M. Busso, J. Gregory, and P. Kline, “Assessing the Incidence and Efficiency of a Prominent Place Based Policy.”

- Jobs seem to go to mix of zone residents and non-residents.
- Wages rise most among zone residents working in zone.
- No increase in rent. Small changes in demographics. But big increase in housing value.
Before EZ
Camden (inside EZ), New Jersey, 1993
After EZ

Same street in Camden (inside EZ), New Jersey, 2003
Assessing a Prominent Place Based Policy (Busso et al.)

Incidence

- Significant increase in earnings for a poor population

- Negligible cost of living increase but possible windfall gain to homeowners

- Little change in demographic composition but probably not literally the original residents
  - Only 57% of households in same house as 5 years ago

- Risk of gentrification and landlord capture over longer run.
  - How to define success?
Assessing a Prominent Place Based Policy (Busso et al.)

Efficiency

- While population response negligible ($\eta_{pop} \approx .15$), quantity being subsidized is local jobs
- Very small target group (unbeknownst to HUD!)

<table>
<thead>
<tr>
<th>TABLE 10—WELFARE ANALYSIS</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total workers/people/households</td>
</tr>
<tr>
<td>Zone residents working in zone</td>
</tr>
<tr>
<td>Zone residents working outside zone</td>
</tr>
<tr>
<td>Nonresidents working in zone</td>
</tr>
<tr>
<td>House renters in the zone</td>
</tr>
<tr>
<td>House owners in the zone</td>
</tr>
</tbody>
</table>

- BGK estimate elasticity of (covered) jobs wrt $(1 - \tau)$ of

  $$\eta_{jobs} \approx 1.25$$

- Efficiency cost roughly 13% of dollar value of subsidy
Efficiency costs depends on what is targeted

Bigger geographic areas not always better
  - Isolated / depressed neighborhoods may be capable of being stimulated without inducing a flood of entrants

Conditionality in benefits
  - Benefits for living and working in area?
  - Benefits tied to residence at some prior date?

A precarious balance: too much stimulus raises cost of living, leads to turnover / gentrification
Reach the intended populations
  - Place itself as an additional dimension of disadvantage?

Entail the smallest efficiency costs
  - More distortionary to influence location or labor supply choices?

But we’ve been assuming behavioral responses are distortionary. Could PBP improve efficiency?
Local Government Spending
Who benefits from government spending in the long run?

(And could place based policies improve efficiency?)

Results are important for setting spending levels and distributing funds across localities

Contributes to literature with by estimating long-run spending effects and workers’ valuation of government services
Figure 1: Supply and Demand Components of a Government Spending Shock

Wage

\[ S_0(w, F_0) \]

\[ S_2(w, F_1) \]

\[ D_1(w, F_1) \]

\[ D_0(w, F_0) \]

Employment
Decomposition of a 1% Increase in Government Spending

Skilled: Supply Shift explains 19% of $\Delta N^S_c$ and 32% of $\Delta w^S_c$

Unskilled: Supply Shift explains 53% of $\Delta N^U_c$ and 46% of $\Delta w^U_c$
Methods: Policy Experiment #1

- Analyze impact of increasing spending per-adult by $1,000
- Median spending per-adult is $10,235
- Change in worker utility is given by

\[
\frac{dV^i}{dv^i} \frac{1}{\lambda^i_c} = N^i_c \frac{dv^i}{\lambda^i_c} \\
= N^i_c \left( dw^i_c + dt^i_c - dr^i_c + \phi^i(w^i_c + t^i_c) \frac{dGS_c}{GS_c} \right)
\]
<table>
<thead>
<tr>
<th>Welfare Effects</th>
<th>Zero Value for Government Services</th>
<th>Including Value for Government Services</th>
</tr>
</thead>
<tbody>
<tr>
<td>Skilled Worker (25%)</td>
<td>$363</td>
<td>$1,012</td>
</tr>
<tr>
<td>Unskilled Worker (25%)</td>
<td>-$92</td>
<td>$751</td>
</tr>
<tr>
<td>Owners of Housing</td>
<td>$325</td>
<td>$325</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Budget Impacts</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Decrease in Transfers</td>
<td>$15</td>
<td>$15</td>
</tr>
<tr>
<td>Increase in Taxes</td>
<td>$290</td>
<td>$290</td>
</tr>
</tbody>
</table>

| Social Welfare                  | $650                              | $1,445                                 |

- An additional $1 of spending raises welfare by $1.45
- Ballard et al. (1985) report MCPF between 1.17 and 1.33
The Incidence of Government Spending (Suárez Serrato and Wingender)

Contribution

- Estimate long-term impacts of government spending
  - Find persistent effects on wages and migration

- Estimate incidence of government spending by skill
  - Supply components of shock explains large mobility responses of the unskilled and lower wage outcomes
  - Incidence on workers may be large enough to motivate spending on utilitarian grounds
  - Heterogenous valuations of government services suggest distribution of funds should target areas with low skill-shares
Moving to Opportunity
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

- Substantial disparities in economic outcomes across low vs. high poverty neighborhoods [e.g., Wilson 1987, Jencks and Mayer 1990, Cutler and Glaeser 1997]

- These disparities motivated the HUD Moving to Opportunity (MTO) experiment in the mid 1990’s
  - Offered a randomly selected subset of families living in high-poverty housing projects housing vouchers to move to lower-poverty areas

- Large literature on MTO has found significant effects on adult health and subjective well-being

- But these studies have consistently found that the MTO treatments had no impact on earnings or employment rates of adults and older youth [e.g. Katz, Kling, and Liebman 2001, Oreopoulos 2003, Sanbonmatsu et al. 2011]

Source: Chetty Hendren Katz (AER, 2016)
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

- We revisit the MTO experiment and focus on its impacts on children who were young when their families moved to better neighborhoods

- Re-analysis motivated by a companion paper that presents quasi-experimental evidence on neighborhood effects [Chetty and Hendren 2015]

  - Key finding: childhood exposure effects

  - Every year in a better area during childhood $\rightarrow$ better outcomes in adulthood

  - Implies that gains from moving to a better area are larger for children who move when young

Source: Chetty Hendren Katz (AER, 2016)
In light of this evidence on childhood exposure effects, we returned to MTO data to examine treatment effects on young children.

Link MTO data to tax data to analyze effects of MTO treatments on children’s outcomes in adulthood.

Children we study were not old enough to observe outcomes in adulthood at the time of the MTO Final Impacts Evaluation (which used data up to 2008).

Source: Chetty Hendren Katz (AER, 2016)
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

- HUD Moving to Opportunity Experiment implemented from 1994-1998
- 4,600 families at 5 sites: Baltimore, Boston, Chicago, LA, New York
- Families randomly assigned to one of three groups:
  1. Experimental: housing vouchers restricted to low-poverty (<10%) Census tracts
  2. Section 8: conventional housing vouchers, no restrictions
  3. Control: public housing in high-poverty (50% at baseline) areas

Source: Chetty Hendren Katz (AER, 2016)
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

Most Common MTO Residential Locations in New York

Experimental
Wakefield
Bronx

Section 8
Soundview
Bronx

Control
King Towers
Harlem
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

- MTO data obtained from HUD
  - 4,604 households and 15,892 individuals
  - Primary focus: 8,603 children born in or before 1991
- Link MTO data to federal income tax returns from 1996-2012
  - Approximately 85% of children matched
  - Match rates do not differ significantly across treatment groups
  - Baseline covariates balanced across treatment groups in matched data

Source: Chetty Hendren Katz (AER, 2016)
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

- We replicate standard regression specifications used in prior work [Kling, Katz, Liebman 2007]

\[ y_i = \alpha + \beta_E^{ITT} \text{Exp}_i + \beta_S^{ITT} \text{S8}_i + s_i \delta_s + \epsilon_i \]

- These intent-to-treat (ITT) estimates identify effect of being offered a voucher to move through MTO

- Estimate treatment-on-treated (TOT) effects using 2SLS, instrumenting for voucher takeup with treatment indicators
  - Experimental take-up: 48% for young children, 40% for older children
  - Section 8 take-up: 65.8% for young children, 55% for older children
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

**Impacts of MTO on Children Below Age 13 at Random Assignment**

(a) Mean Poverty Rate in Tract (ITT)
Post RA to Age 18

<table>
<thead>
<tr>
<th>Group</th>
<th>Mean Poverty Rate (%)</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Control</td>
<td>41.2%</td>
<td>0.0001</td>
</tr>
<tr>
<td>Section</td>
<td>33.2%</td>
<td>0.0001</td>
</tr>
<tr>
<td>Experimental</td>
<td>30.9%</td>
<td>0.0001</td>
</tr>
</tbody>
</table>

(b) Mean Poverty Rate in Tract (TOT)
Post RA to Age 18

<table>
<thead>
<tr>
<th>Group</th>
<th>Mean Poverty Rate (%)</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Control</td>
<td>41.2%</td>
<td>0.0001</td>
</tr>
<tr>
<td>Section</td>
<td>29.1%</td>
<td>0.0001</td>
</tr>
<tr>
<td>Experimental</td>
<td>19.6%</td>
<td>0.0001</td>
</tr>
</tbody>
</table>
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

Impacts of MTO on Children Age 13-18 at Random Assignment

(a) Mean Poverty Rate in Tract (ITT)
Post RA to Age 18

- Control: 47.9%
- Section 8: 39.3%
- Experimental: 37.9%

(b) Mean Poverty Rate in Tract (TOT)
Post RA to Age 18

- Control: 47.9%
- Section 8: 32.5%
- Experimental: 23.2%

*p = 0.0001 for all groups
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

Impacts of MTO on Children Below Age 13 at Random Assignment

(a) Individual Earnings (ITT)
(b) Individual Earnings (TOT)

<table>
<thead>
<tr>
<th>Group</th>
<th>Individual Income at Age ≥ 24 ($)</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Control</td>
<td>11,270</td>
<td>0.101</td>
</tr>
<tr>
<td>Section 8</td>
<td>12,380</td>
<td>0.014</td>
</tr>
<tr>
<td>Experimental</td>
<td>12,894</td>
<td>0.014</td>
</tr>
<tr>
<td>Control</td>
<td>11,270</td>
<td>0.101</td>
</tr>
<tr>
<td>Section 8</td>
<td>12,994</td>
<td>0.014</td>
</tr>
<tr>
<td>Experimental</td>
<td>14,747</td>
<td>0.014</td>
</tr>
</tbody>
</table>
Chetty Hendren Katz (AER, 2016) on MTO
Effect of Exposure to Better Neighborhoods

Impacts of Experimental Voucher by Age of Earnings Measurement
For Children Below Age 13 at Random Assignment

Experimental Vs. Control ITT on Earnings ($)
-1000 0 1000 2000 3000
20 21 22 23 24 25 26 27 28
Could improving places eventually save money?

**Exposure specification:** extra year of good neighborhood → extra $566 of age 26 earnings!

**TABLE 8**

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Experimental × Age at RA</td>
<td>-364.1* (199.5)</td>
<td>-723.7** (256.5)</td>
<td>-564.9* (282.8)</td>
<td>-171.0** (55.16)</td>
<td>-0.582* (0.290)</td>
<td>0.261* (0.139)</td>
<td>-65.81** (23.88)</td>
</tr>
<tr>
<td>Section 8 × Age at RA</td>
<td>-229.5 (208.9)</td>
<td>-338.0 (266.4)</td>
<td>157.2 (302.0)</td>
<td>-117.1* (63.95)</td>
<td>-0.433 (0.316)</td>
<td>0.0109 (0.156)</td>
<td>-42.48* (24.85)</td>
</tr>
<tr>
<td>Experimental</td>
<td>4823.3* (2404.3)</td>
<td>9441.1** (3035.8)</td>
<td>8057.1* (3760.9)</td>
<td>1951.3** (575.1)</td>
<td>8.309* (3.445)</td>
<td>-4.371* (1.770)</td>
<td>831.2** (279.4)</td>
</tr>
<tr>
<td>Section 8</td>
<td>2759.9 (2506.1)</td>
<td>4447.7 (3111.3)</td>
<td>-1194.0 (3868.2)</td>
<td>1461.1* (673.6)</td>
<td>7.193* (3.779)</td>
<td>-1.237 (2.021)</td>
<td>521.7* (287.5)</td>
</tr>
<tr>
<td>Number of Observations</td>
<td>20043</td>
<td>20043</td>
<td>3956</td>
<td>20127</td>
<td>20043</td>
<td>15798</td>
<td>20043</td>
</tr>
<tr>
<td>Control Group Mean</td>
<td>13807.1</td>
<td>16259.9</td>
<td>14692.6</td>
<td>21085.1</td>
<td>6.6</td>
<td>23.7</td>
<td>627.8</td>
</tr>
</tbody>
</table>

**Source:** Chetty, Hendren, Katz (2015)
Moved to Opportunity
How does growing up in a disadvantaged neighborhood affect long-run child outcomes?

- Large observational literature shows children from disadvantaged areas have notably worse outcomes
  - Ellen and Turner (1997); Cutler and Glaeser (1997); Altonji and Mansfield (2014); Chetty et al., (2014)

- Yet, some experimental evidence finds few significant effects of moving to better neighborhoods
  - Katz et al. (2001); Oreopolous (2003); Sanbonmatsu et al., (2011)

- Existence and size of neighborhood effects is uncertain
  - This view has started to change due to recent work: Chetty, Hendren and Katz (2015); Chetty and Hendren (2015)

Source: Chen (2017)
What is public housing?

- Goal: Provide “decent” housing for low-income families
- Large residential buildings (high-rises) built in close proximity
  - A collection of buildings is called a housing project

Source: Chen (2017)
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

Figure 1: Robert Taylor Homes
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

- Federally supported program, but owned and operated by local (city) authority
- Assistance is *not* an entitlement – long waiting lists
- Value of subsidy is large: $8,000 per year (HUD, 2015)

Source: Chen (2017)
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

Descriptive statistics:
1. Third largest public housing system during the 1990s
2. Average household income: $7,000
3. 20% of units have more than 5 people
4. Nearly all residents are African-American

Source: Chen (2017)
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

- Reaction to serious management and infrastructure problems
  - Buildings built during the 50s and 60s cheaply
  - Few believed the Chicago Housing Authority (CHA) could deal with maintenance issues
    - Scandals revealed officials had mismanaged public funds
  - Local politicians proposed demolition and expanding voucher assistance
    - Limited funding for demolition

Source: Chen (2017)
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

- Limited funding $\Rightarrow$ selection of buildings based on specific maintenance issues (Jacob, 2004)
  - Initial demolitions motivated by specific crises
  - Ex. Pipes burst in Robert Taylor high-rise buildings

Source: Chen (2017)
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

- Provided housing vouchers and compensated for moving costs
  - Note: Vouchers and project-based assistance have the same rules ⇒ No effect on budget set

- Households moved to lower poverty areas:

![Densities of Neighborhood Poverty Rates](image)
Moved to Opportunity (Chen, 2017)

Contribution

- Provide new evidence on neighborhood effects for children from two different housing policy interventions

  1. Natural experiment created by public housing demolition
  2. Housing voucher lottery

- Compare these two contexts to answer two questions:

  1. What are the benefits of relocating youth in a general population?
  2. Do children of volunteers benefit more or less than average?

Source: Chen (2017)
Natural Experiment Research Design
Public Housing Demolition in Chicago

- Identification: Displacement unrelated to resident characteristics (Jacob, 2004)

- Results: Displaced children are (1) more likely to work, (2) have higher annual earnings and (3) have fewer arrests for violent crime
Moved to Opportunity (Chen, 2017)
The Long-Run Effect of Public Housing Demolition on Labor Market Outcomes of Children

Lottery Design
The 1997 Chicago Housing Voucher Lottery

- Main finding: Small and not statistically significant effects on lottery children outcomes

Source: Chen (2017)
Notable contrast between demolition and lottery results

Comparing Employment Effects Across Experiments

Impact on Labor Market Participation

- Demolition (Chyn 2015)
- Demolition Re-weighted (Chyn 2015)
- CHAC 1997 (Chyn 2015)
Interpreting the Evidence and Implications

- Pattern consistent with larger benefits for children from households where parents have low demand for moving
  - Demolition $\Rightarrow$ General set of households
  - Lottery $\Rightarrow$ Subset with high willingness to move

- “Reverse Roy” and parental behavior in education studies:
  - Boston charter schools (Walters, 2015)
  - North Carolina school choice (Hastings et al., 2008)

- Benefits to moving children from public housing may be larger than estimates based on experiments such as MTO
Million Dollar Plants
Here there are two stable equilibria: one much better than the other.
An un-natural experiment

Source: Davis and Weinstein (2002)
What is the impact of the opening of a large manufacturing plant on the total factor productivity (TFP) of incumbent plants in the same county?

This work contributes to the policy debate on the importance of location-based incentives.

The authors add to the literature by providing evidence for the existence of agglomeration spillovers in a specific industry.
Predictions in case of positive spillovers:

- The opening of a new plant will increase TFP of incumbents
- The increase in TFP may be larger for firms that are economically “closer” to new plant
- The density of economic activity in the county will increase as firms move in
- The price of locally supplied factors of production will increase
Empirical strategy:

- Comparing the “winning” counties (where the new plant is located) to the “losing” ones (runner-ups) allows to isolate the effects that result solely from agglomeration.

- Identification: use location rankings of firms to identify a valid counterfactual for what would have happened to incumbent plants in “winning” counties in the absence of the plant opening.

- The research design is convincing at testing for agglomeration - it is realistic that “winning” counties would benefit from the concentration of economic activity.
“Million Dollar Plant” articles from the Site Selection list the “winning” and “losing” counties.

Information about the plants comes from the Census Bureau’s Standard Statistical Establishment List (SSEL), the Annual Survey of Manufactures (ASM) and the Census of Manufactures (CM).

The data on plant variables such as employment and value of shipments is panel for the opening year + 8 years before.
Figure: Incumbents’ Productivity in Winning vs Losing Counties (Greenstone et al.)
Local Economic Development: TVA
What are the effects of the Tennessee Valley Authority policy on local economies?

Paper informs the debate on spatially targeted policies

Kline and Moretti are the first to empirically quantify the long run social costs and benefits of a place based policy.
The empirical strategy is to compare long run changes in TVA counties with long run changes in non-TVA counties with similar characteristics.

This allows to isolate the effects of the TVA policy on economic growth, controlling for other influences.

Regression model: \( y_{it} - y_{it-1} = \alpha + \beta X_i + (\epsilon_{it} - \epsilon_{it-1}) \)

- \( y_{it} - y_{it-1} \) is the change in the dependent variable between years \( t - 1 \) and \( t \) for county \( i \).
- \( X_i \) is the vector of preprogram characteristics.
The data comes from the Population Census, the Manufacturing Census, the Agricultural Census, and from Fishback, Haines, and Kantor (2011).

It is used to create a county-level panel from 1900 to 2000.

Some of the variables are imprecise, and substantial measurement error is likely to be present at the beginning of the sample period.
Figure: Impact of TVA on Growth Rate (Kline and Moretti)

<table>
<thead>
<tr>
<th>Outcome</th>
<th>Point Estimate (Unadjusted)</th>
<th>Clustered S.E.</th>
<th>Point Estimate (Controls)</th>
<th>Clustered S.E.</th>
<th>Spatial HAC</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1) Population</td>
<td>0.007</td>
<td>(0.016)</td>
<td>0.010</td>
<td>(0.012)</td>
<td>(0.012)</td>
<td>1776</td>
</tr>
<tr>
<td>(2) Total Employment</td>
<td>-0.009</td>
<td>(0.016)</td>
<td>0.005</td>
<td>(0.013)</td>
<td>(0.013)</td>
<td>1776</td>
</tr>
<tr>
<td>(3) Housing Units</td>
<td>-0.006</td>
<td>(0.015)</td>
<td>0.007</td>
<td>(0.011)</td>
<td>(0.011)</td>
<td>1776</td>
</tr>
<tr>
<td>(4) Average Manufacturing Wage</td>
<td>0.009</td>
<td>(0.018)</td>
<td>0.010</td>
<td>(0.021)</td>
<td>(0.021)</td>
<td>1428</td>
</tr>
<tr>
<td>(5) Manufacturing Share</td>
<td>0.007*</td>
<td>(0.004)</td>
<td>0.005</td>
<td>(0.004)</td>
<td>(0.004)</td>
<td>1776</td>
</tr>
<tr>
<td>(6) Agricultural Share</td>
<td>-0.007*</td>
<td>(0.004)</td>
<td>-0.001</td>
<td>(0.005)</td>
<td>(0.005)</td>
<td>1776</td>
</tr>
<tr>
<td>(7) Average Agricultural Land Value</td>
<td>0.078***</td>
<td>(0.021)</td>
<td>0.025</td>
<td>(0.018)</td>
<td>(0.018)</td>
<td>1746</td>
</tr>
</tbody>
</table>
**Figure:** Impact of TVA on Growth Rate (Kline and Moretti)

<table>
<thead>
<tr>
<th>Outcome</th>
<th>Point Estimate (Unadjusted)</th>
<th>Clusted S.E.</th>
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<th>Clustered S.E.</th>
<th>Spatial HAC</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1) Population</td>
<td>0.004</td>
<td>(0.021)</td>
<td>0.007</td>
<td>(0.020)</td>
<td>(0.018)</td>
<td>1907</td>
</tr>
<tr>
<td>(2) Average Manufacturing Wage</td>
<td>0.027***</td>
<td>(0.006)</td>
<td>0.005</td>
<td>(0.004)</td>
<td>(0.005)</td>
<td>1172</td>
</tr>
<tr>
<td>(3) Agricultural Employment</td>
<td>-0.130***</td>
<td>(0.026)</td>
<td>-0.056**</td>
<td>(0.024)</td>
<td>(0.027)</td>
<td>1907</td>
</tr>
<tr>
<td>(4) Manufacturing Employment</td>
<td>0.076***</td>
<td>(0.013)</td>
<td>0.059***</td>
<td>(0.015)</td>
<td>(0.023)</td>
<td>1907</td>
</tr>
<tr>
<td>(5) Value of Farm Production</td>
<td>-0.028</td>
<td>(0.028)</td>
<td>0.002</td>
<td>(0.032)</td>
<td>(0.026)</td>
<td>1903</td>
</tr>
<tr>
<td>(6) Median Family Income (1950-2000 only)</td>
<td>0.072***</td>
<td>(0.014)</td>
<td>0.021</td>
<td>(0.013)</td>
<td>(0.011)</td>
<td>1903</td>
</tr>
<tr>
<td>(7) Average Agricultural Land Value</td>
<td>0.056***</td>
<td>(0.013)</td>
<td>-0.002</td>
<td>(0.012)</td>
<td>(0.016)</td>
<td>1906</td>
</tr>
<tr>
<td>(8) Median Housing Value</td>
<td>0.040**</td>
<td>(0.017)</td>
<td>0.005</td>
<td>(0.015)</td>
<td>(0.015)</td>
<td>1906</td>
</tr>
</tbody>
</table>
Formal Model:

- Utility is equalized across counties in each year: \( \ln w_{it} + M_{it} = \bar{u}_t \)

- It is used to create a county-level panel from 1900 to 2000

- Production function: \( Y_{it} = A_{it} K_{it}^{\alpha} F_i^{\beta} L_{it}^{1-\alpha-\beta} \)

- \( A_{it} \) is a local productivity level, \( L_{it} \) is the number of manufacturing workers, \( K_{it} \) is the capital stock, \( F_i \) is a fixed nonreproducible factor (i.e. natural features)
Agglomeration Economies and the Tennessee Valley Authority (Kline and Moretti)

Methods: Formal Model

- Labor demand: \( \ln w_{it} = C - \frac{\beta}{1 - \alpha} \ln L_{it} + \frac{\beta}{1 - \alpha} \ln F_i - \frac{\alpha}{1 - \alpha} \ln r_t + \frac{1}{1 - \alpha} \ln A_{it} \)

- \( C \equiv \ln(1 - \alpha - \beta) + \frac{\alpha}{1 - \alpha} \ln \alpha \)

- \( \ln A_{it} \) can be decomposed into a locational advantage component, a component due to agglomeration effects, an effect of TVA, and an idiosyncratic component:
  \( \ln A_{it} = g\left(\frac{L_{it} - 1}{R_i}\right) + \delta_t D_i + \eta_i + \gamma_t + \varepsilon_{it} \)

  - \( D_i \) is a dummy for TVA exposure
Agglomeration Economies and the Tennessee Valley Authority (Kline and Moretti)

Methods: Formal Model

- Direct TVA effect: impact on public infrastructure, as captured by $\delta_t$ coefficients

- Indirect TVA effect: increases in employment may cause further increases in productivity (agglomeration)

- The impact of a marginal increase in the productivity of TVA’s investments on output:
  \[
  \frac{dY_i}{d\delta} = \frac{1}{1-\alpha} Y_i (D_i + \frac{1-\alpha-\beta+\sigma_i}{L_i} dL_i)
  \]

  - $\sigma_i$ is the local agglomeration elasticity

- Steady state productivity: $\ln A_i = g\left(\frac{L_i}{R_i}\right) + \eta_i + \delta D_i$
Agglomeration Economies and the Tennessee Valley Authority (Kline and Moretti)

Methods: Structural Estimation

Structural Estimation:

\[
\ln(L_{it}) - \ln(L_{it-1}) = \frac{1 - \alpha}{\beta} (\ln w_{it} - \ln w_{it-1}) + \frac{\delta_t - \delta_{t-1}}{\beta} D_i \\
+ \frac{\theta_1}{\beta} [g_1(L_{it-1}) - g_1(L_{it-2})] + \frac{\theta_2}{\beta} [g_2(L_{it-1}) - g_2(L_{it-2})] \\
- g_2(L_{it-2}) R_i + \frac{\theta_3}{\beta} [g_3(L_{it-1}) - g_3(L_{it-2})] R_i \\
+ X_i'\tilde{\lambda} + \tilde{\gamma}_t + \gamma_{t-1} + \tilde{v}_{it}
\]

- \(\frac{\delta_t - \delta_{t-1}}{\beta}\) gives the change in direct effects of TVA between decades

- Spline coefficients \(\frac{\theta_i}{\beta}\) determine the indirect effects, since they give the labor demand effects of within the relevant density range
**Table 6: Structural Estimates of Agglomeration Function (log basis)**

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
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<tbody>
<tr>
<td><strong>OLS</strong></td>
<td>OLS</td>
<td>OLS</td>
<td>OLS</td>
<td>2SLS</td>
<td>2SLS</td>
<td>2SLS</td>
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<tr>
<td><strong>Change in Log Manufacturing Density Spline Components:</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
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<tr>
<td><strong>Low</strong></td>
<td>0.078</td>
<td>0.053</td>
<td>0.052</td>
<td>0.349</td>
<td>0.323</td>
<td>0.325</td>
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<tr>
<td></td>
<td>(0.030)</td>
<td>(0.031)</td>
<td>(0.030)</td>
<td>(0.109)</td>
<td>(0.122)</td>
<td>(0.123)</td>
</tr>
<tr>
<td></td>
<td>[182.83]</td>
<td>[149.61]</td>
<td>[148.34]</td>
<td></td>
<td></td>
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<tr>
<td><strong>Medium</strong></td>
<td>0.072</td>
<td>0.075</td>
<td>0.069</td>
<td>0.339</td>
<td>0.327</td>
<td>0.319</td>
</tr>
<tr>
<td></td>
<td>(0.049)</td>
<td>(0.050)</td>
<td>(0.050)</td>
<td>(0.097)</td>
<td>(0.101)</td>
<td>(0.103)</td>
</tr>
<tr>
<td></td>
<td>[92.69]</td>
<td>[96.61]</td>
<td>[97.01]</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>High</strong></td>
<td>0.084</td>
<td>0.090</td>
<td>0.086</td>
<td>0.306</td>
<td>0.304</td>
<td>0.307</td>
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<tr>
<td></td>
<td>(0.059)</td>
<td>(0.059)</td>
<td>(0.059)</td>
<td>(0.134)</td>
<td>(0.135)</td>
<td>(0.136)</td>
</tr>
<tr>
<td></td>
<td>[206.26]</td>
<td>[204.81]</td>
<td>[202.69]</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Log Manufacturing Wages</strong></td>
<td><strong>-1.5</strong></td>
<td><strong>-1.5</strong></td>
<td><strong>-1.5</strong></td>
<td><strong>-1.5</strong></td>
<td><strong>-1.5</strong></td>
<td><strong>-1.5</strong></td>
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<tr>
<td><strong>TVA</strong></td>
<td>0.024</td>
<td>0.027</td>
<td>0.029</td>
<td>0.008</td>
<td>0.011</td>
<td>0.012</td>
</tr>
<tr>
<td></td>
<td>(0.013)</td>
<td>(0.013)</td>
<td>(0.014)</td>
<td>(0.011)</td>
<td>(0.011)</td>
<td>(0.012)</td>
</tr>
<tr>
<td><strong>Regional Trends</strong></td>
<td>no</td>
<td>no</td>
<td>yes</td>
<td>no</td>
<td>no</td>
<td>yes</td>
</tr>
<tr>
<td><strong>1940 Manufacturing Density</strong></td>
<td>no</td>
<td>yes</td>
<td>yes</td>
<td>no</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td><strong>Decade Effects</strong></td>
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<td>yes</td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td><strong>Controls for 1920 and 1930 characteristics</strong></td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td><strong>P-value equal slopes</strong></td>
<td>0.981</td>
<td>0.799</td>
<td>0.837</td>
<td>0.891</td>
<td>0.980</td>
<td>0.982</td>
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<tr>
<td><strong>P-value slopes equal zero</strong></td>
<td>0.039</td>
<td>0.141</td>
<td>0.173</td>
<td>0.002</td>
<td>0.007</td>
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<tr>
<td><strong>N</strong></td>
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<td>5462</td>
<td>5318</td>
<td>5318</td>
<td>5318</td>
</tr>
</tbody>
</table>
Other considerations: Second best arguments

Correct prior distortions that can interact w/ place:

- Deductibility of state and local taxes (Albouy, 2008)

- Hiring costs (Kline and Moretti, 2013)

- State sales / business taxes (Fajgelbaum, Morales, Suarez Serrato, Zidar, 2016)

- Housing regulations (Hsieh and Moretti, 2016)

- Payroll taxes?
Closing thoughts

- Place conveys useful information about preferences and endowments
- Odd to ignore when setting policy
- Equity - efficiency tradeoff looms large but “triangle” view may miss forest for trees
- Some under-explored questions:
  1. Picking winners: what do economists have to offer?
  2. Paternalism and place: nudge households to move?
  3. Coordinating expectations: is economic development like faith healing?
Appendix: Discrete Choice
Aside on Discrete Choice

- Brief review of discrete choice
- CDF of tastes and demand curves
- Link to demand elasticities
- See Ken Train’s *Discrete Choice Methods with Simulation* (free online) for very clear, helpful discussion
Consumers decide whether or not to buy
Consumers decide whether or not to buy
Consumers decide whether or not to buy

- The first graph shows the share of consumers buying a product is 50% when it’s price is $5.
- The second graph shows the share of consumers buying a product is 30% when it’s price is $6.
- How can we think about how responsive demand will be to changes in price when consumers are making discrete (i.e., buy or not) choices?
Suppose that individual \( i \) buys if her value exceeds the price, i.e., buy if \( v_i > P \)

This value can be a function of common things (e.g., income, credit conditions, etc) or idiosyncratic tastes but at this stage, specifying what is in \( v_i \) doesn’t matter. The fraction of people who buy is:

\[
\text{Prob}(Q = 1) = P(v_i > P) = 1 - F(P)
\]

where \( F(x) \) is the c.d.f. of \( v_i \). Note this is why the demand curve looks like a CDF rotated clockwise 90 degrees.

A c.d.f. describes the probability that a real-valued random variable \( X \) with a given probability distribution will be found to have a value less than or equal to \( x \).
What is the elasticity of this curve?

\[ Q(P) = N(1 - F(P)) \]  

(7)

where \( N \) is the size of the population (e.g., number of potential consumers in your market)

\[ \varepsilon^D = \frac{dQ(P)}{dP} \frac{Q}{P} \]  

(8)
What is the derivative?

\[
\frac{dQ(P)}{dP} = -Nf(P)
\]  \hspace{1cm} (9)

- where \( N \) is the size of the population (e.g., first time home buyers in an area)
- \( f(x) \) is the probability density function (p.d.f.)
Elasticity of Demand

\[ \varepsilon^D = \frac{dQ(P)}{dP} \frac{P}{Q} \]  

\[ = -N f(P) \frac{P}{N(1 - F(P))} \]  

\[ = -f(P) \frac{P}{1 - F(P)} \]  

- What matters for responsiveness is how big the density is at \( P \) relative to 1 minus the CDF
From $5, a $1 dollar increase in price \downarrow\ demand by 20%
From $8, a $1 dollar increase in price $\downarrow$ demand by 2%
Elasticity of Demand: In words

Takeaways:

- For very homogeneous populations, you’ll have very elastic demand
- If tastes are more spread out, you’ll see smaller responses
- At the extreme in which everyone is the same, demand will be a step function, so there is some price above which no one will buy and below which everyone will buy.
- In this case, things will be very inelastic at high prices, but very elastic near the price, and then unresponsive at very low prices
- Thinking about consumer choice in this way will be helpful for evaluating how effective sales can be