Task #1 Task #2

# DATA SCIENTIST ASSESSMENT DATA LAB AT PHILADELPHIA DAO

J.J. Naddeo

Georgetown University and Justice Innovation Lab

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# Task #1: Research Design

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Task #2

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- Which research design(s) might be appropriate to understand the causal relationship (if any) between housing improvements and shootings?
- What method(s) would you consider or plan to use to investigate these questions?
- What limitations do you foresee in such an analysis? Are there complementary research designs that could be conducted to mitigate them?

#### RESEARCH DESIGN: RCT

- The cleanest (and most expensive) way to answer such questions is to run a randomized experiment
- Here if possible one would randomly pick neighborhoods and then "treat" them with housing improvements
  - Increased lighting, green space, renovations of abandoned houses, re-pavement of roads/sidewalks, etc.
- Method: Use OLS to evaluate if treatment reduces shootings in neighborhoods that receive treatment

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- Imbalance by observables
  - Even with randomization there can be imbalance (e.g. treatment neighborhoods are higher income)
  - Solution: Control for observables and/or block randomize using some observables that are first-order important

### Research Design: Observational

- If RCT is not feasible one may try to use historical observational data to determine causality
- First look for events that may have changed the "quality" of housing in neighborhoods but did not impact any other variables that could affect shootings
- Example: Citywide renovation of public goods that was rolled out in a staggered manner
- Method: Difference in differences
  - Compare shootings in treated areas to those yet treated
  - Be careful and use correct comparisons (Roth et al. 2023)
- Limitations: Parallel trends could be impacted by spillover effect previously discussed

# Task #2: Theory $\rightarrow$ Practice

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  - Distribution of individuals on road
  - Distribution of contraband/criminal activity

## Task #2: Theory $\rightarrow$ Practice

- Testing for racial bias in stop/search behavior is difficult
  - Distribution of individuals on road
  - Distribution of contraband/criminal activity
- Potential solutions:
  - Look at differences in hit rates
  - Look at differences in search/frisk rates

#### HIT RATES

- "Hit rates" by race,  $\gamma^r$
- If  $\gamma^A > \gamma^B$  for races A and B it suggests that there is racial bias against group B
- Restrict to vehicle stops in 2021 in districts with at least 100 stops
- Mean hit rate in sample: 2.4%

#### HIT RATES

- "Hit rates" by race,  $\gamma^r$
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- Restrict to vehicle stops in 2021 in districts with at least 100 stops
- Mean hit rate in sample: 2.4%
- Compared to vehicles driven by White (non-latino) motorists vehicle searches of Black (non-Latino) [-5.2 p.p.], White Latino [-7 p.p.] drivers are less successful and searches of Asian [7.7 p.p.] are more sucessful
  - controls: age, gender, month, day of week, and district the stop occurred in

# SEARCH/FRISK RATES

- Can also look at rates of search/frisk
- Restrict to vehicle stops in 2021 in districts with at least 100 stops
- Mean search/frisk rate in entire sample 13.4%
- Results indicate that compared to vehicles driven by White (non-Latino) motorists vehicle are searched at higher rates for Black (non-Latino) [3.6 p.p.] and White - Latino [0.2 p.p. NSS], lower rates for Asian [-2.6 p.p.] drivers
  - controls: age, gender, month, day of week, and district the stop occurred in

## Veil of Darkness

- Using variation in light we can look at how these relationships vary by visibility
- The intuition is that darkness shields police from seeing the race/ethnicity of drivers and makes stops less susceptible to bias
- Comparing night to day could introduce issues with driving patterns→limit to 3 hours where transition occurs
  - Note: should do a more robust analysis using daylight savings and controlling for hour as in Grogger and Ridgeway (2006), but did not have time
- Define dark as when sun is 12° below horizon
- Hit rate: Racial differences in hit rates closes under veil of darkness
- Search/frisk rates: Gap widens for Black (non-Latino) and Asian

#### LIMITATIONS

- Inframarginality bias: For hit rate analysis we assume that we are identifying the marginal search, could be the case we aren't (Simoiu et al. 2017)
- Misreporting: Police are likely aware of hit rate analysis and may misreport drivers race/ethnicity (Luh 2022)
- Street lighting could make veil of darkness ineffective, especially in well lit urban areas (Horrace and Rohlin 2016)
- Darkness may change traffic patterns which changes distribution of drivers/risk

#### FUTURE WORK

- Incorporate information that control for incidents
- Merge in information regarding streetlight location/brightness
- Conduct an regression discontinuity in time (potentially using a donut estimator) around light/dark
- Explore heterogeneity in