Washington University in St. Louis

Responsive Politics or Static Cities?

Are local governments responsive given their overlapping nature? To examine responsiveness at this level, I **implement** a framework that takes into account multiple overlapping institutions and estimate a novel measure of local preferences for cities over time. I have three major findings using a within-between random-effects model: (1) cross-sectional responsiveness exists; (2) I find mixed evidence for dynamic responsiveness; (3) I find suggestive evidence that consolidated governance fosters greater responsiveness. In all, I reframe the responsiveness discussion away from a single governing unit to a holistic system of overlapping institutions.

Motivation: Overlapping Institutions of Cincinnati, OH



Research Questions

- RQ1: Does cross-sectional responsiveness exist?
- RQ2: Does dynamic responsiveness exist?
- RQ3: Does consolidated government modify the effect of public opinion on policy outcomes?

Sample: Fiscally Standardized Cities

- Fiscally Standardized Cities (FiSC) provide an aggregate measure of taxation and spending of overlapping governments within the geographic boundaries of cities (Langley 2013)
- Sample includes 200 cities across 10 years
- Traditional measure of expenditures and tax revenue fail to capture the full range of goods, services, and costs of local government
- The FiSC measure of government provides a better approach to analyze local governance



https://bryantjmoy.github.io

Responsiveness in a Fragmented Local Politics

Bryant J. Moy

OTR Special District —Downtown Special District

Stacked Expenditures and Tax Revenue for FiSC: Cincinnati, OH

Novel Measure of Dynamic Local Ideology (N > 500,000)

- New dataset including self-placement ideology from multiple surveys: Cooperative Election Study
- National Annenberg Election Survey
- Gallup Social Series Poll
- Over a half a million respondents
- Dynamic Multilevel Regression and Post-Stratification (MRT) (Gelman et. al. nd.)

 $x_{it} = \gamma_1 \text{year_std} + \gamma_2 \text{year_std_sq} + \alpha_{r[i]}^{\text{race3}} + \alpha_{e[i]}^{\text{education3}} + \alpha_{g[i]}^{\text{gender2}} + \alpha_{l[i]}^{\text{location}} + \alpha_{l[i]}^{\text{location}} + \alpha_{e[i],t[i]}^{\text{gender2}} + \alpha_{g[i],t[i]}^{\text{gender2}} + \alpha_{l[i],t[i]}^{\text{gender2}} + \alpha_{l[i],t[i]}^{\text{$

- x_{it} = Ideology
- *i* = respondents
- t = Year

• *j* = city

t = year

- γ_1 = Year (Standardized) • γ_2 = Year Squared
- (Standardized)
- race3 = White, Black, Other

Estimates of Conservatism Over Time: Cincinnati and Columbus OH



Modeling Strategy: Within-Between Random Effects Model

Mundlak (1978), Bell and Jones (2015):



- \bar{x}_i = Mean of Time-Varying Variables (Between City Effects) • z_i = Level 2: Time-Invariant Variables
- μ_i = Level 2 error (Aggregated unobserved group-level effect) • ϵ_{it} = Level 1 error
- x_{it} = Level 1: Mean Centered Time-Varying Variables (Ideology)

• y_{it} = Total Expenditures, Total Tax Revenue

Advantages of Within-Between Random Effects Model

- Dynamic and Cross-Sectional variation in single model
- Same results as Two-Way Fixed Effects and Pooled Cross-Sectional models
- Ability to include time-invariant features

- - education = No high school, Some College, BA or higher
 - gender = Male, Female
- Iocation = cities (via zipcode)

- (3)



- Median Income Population
- Median Home Value Black Share
- Results less strong using traditional measure of policy outcomes

Effect Modification: Support for RQ 3

A. Between Unit Effects (RQ 1) Intercept

Ave. Conservatism (Cross-section

Consolidated Government

B. Within Unit Effects (RQ 2) Conservatism (Dynamic)

C. Cross-Level Interactions (RQ Conservatism*Consolidated Gov

- Clear evidence of cross-sectional representation in local government
- Evidence that dynamic responsiveness exist for expenditures
- Dynamic responsiveness for tax revenue is inconclusive
- Suggestive evidence that more concentrated governance fosters responsiveness
- [1] Andrew Bell and Kelvyn Jones. Explaining fixed effects: Random effects modeling of time-series cross-sectional and panel data. Political Science Research and Methods, 3(1):133–153, 2015.
- [2] Andrew Gelman, Jeffrey Lax, Justin Phillips, Jonah Gabry, and Robert Trangucci. Using multilevel regression and poststratification to estimate dynamic public opinion. Unpublished, 2019.
- [3] Adam H Langley. Methodology used to create fiscally standardized cities database. Lincoln Institute of Land Policy., 2013.
- [4] Yair Mundlak. On the pooling of time series and cross section data. Econometrica: Journal of the Econometric Society, pages 69–85, 1978.

Main Results: Support for RQ 1; Mixed for RQ 2



Results robust to the inclusion of time-variant and time-invariant covariates:

- Income Inequality
- Consolidated Government

	Expenditures	Taxation
	-2660.89	-3470.26*
	[-7587.77, 2205.63]	[-5629.74, -1288.07]
ional)	-3279.37*	-1125.86*
	[-4808.82, -1742.47]	[-1805.18, -423.13]
	293.89	183.53
	[-416.58, 1023.96]	[-152.88, 514.46]
	-3293.64*	-254.64
	[-5256.02, -1340.93]	[-970.36, 455.89]
3)		
V.	-1712.52*	-657.24*
	[-2672.73, -745.11]	[-1008.77, -301.05]

Takeaways

References