

Applications in Labor Economics: Minimum Wage and Labor Demand

Prof. Tzu-Ting Yang
楊子霆

Institute of Economics, Academia Sinica
中央研究院經濟研究所

April 24, 2023

Minimum Wage and Labor Demand

Theoretical Framework

- Neoclassical economic theory suggests that binding price floor policies, including minimum wages, should lead to a non-market equilibrium
 - Since workers' productivity does not change as minimum wage increases
 - But employers have to hire them with higher wage
 - Excess supply and diminished demand
- Some previous empirical studies have questioned the extent to which this prediction holds in the low wage labor market
 - Many estimates suggesting a negligible impact of higher minimum wages on employment

Minimum Wage and Labor Demand

Taiwan's Facts

- Minimum wage in 2023 is 176 NT\$ per hour or 26,400 NT\$ per month
- Government did not change minimum wage during 1997 to 2007
 - Hourly wage: 66 NT\$ during 1997 to 2007
 - Monthly salary: 15,840 NT\$ during 1997 to 2007
- Since 2010, almost every year government raises minimum wage
 - Hourly wage: 98 NT\$ in 2010
 - Monthly salary: 17,880 NT\$
 - Increase by about 80%

Minimum Wage and Labor Demand

Taiwan's Facts

- Huang and Chou (2019) examines the effect of increasing minimum wage during 2013-2018 on employment
 - Reduce the employment rate by 10%
 - Effects are larger for old workers (above age 55)

Empirical Example: Ekaterina Jardim et. al (2018)

Ekaterina Jardim, Mark C. Long, Robert Plotnick, Emma van Inwegen, Jacob Vigdor, and Hilary Wething (2018) “**Minimum Wage Increases, Wages, and Low-Wage Employment: Evidence from Seattle**” American Economic Journal: Economic Policy, vol. 14(2), pages 263-314

- The authors estimate the causal effect of minimum wage on wages and low-wage employment by using synthetic control method

Empirical Example: Ekaterina Jardim et. al (2018)

Motivation

- They use rich administrative data on employment, earnings, and hours in Washington State to re-examine this prediction
- Seattle increases minimum wage from \$9.47 to as much as \$11 in April 2015
 - Increases again to \$13 in January 2016

Empirical Example: Ekaterina Jardim et. al (2018)

Motivation

- They examine the impact of a minimum wage increase for employment across **all categories of low-wage employees**
- Previous literature cannot observe wage directly
 - Focus on lower-wage industries, such as the restaurant sector, or on lower-productivity employees such as teenagers
 - May yield **attenuated estimates** of the effect as they blend workers for whom the minimum wage is binding with workers for whom it is not

Empirical Example: Ekaterina Jardim et. al (2018)

Motivation

- They examine both the extensive (employment) and intensive margins (hours of worked)
 - Prior studies commonly analyze only measures of “headcount” employment (extensive margin)
 - Ignoring the reality that most low-wage jobs are part-time in nature
 - The intensive margin may be a significant dimension of adjustment

Empirical Example: Ekaterina Jardim et. al (2018)

Policy Background

- In June 2014, the City of Seattle passed a minimum wage ordinance
 - Gradually increased the minimum wage within Seattle's city boundaries to \$15
 - The phase-in rate differed by employer size, and offered some differentiation for employers who pay tips or health benefits
- They study the impact of the 2015 and 2016 minimum wage increases in Seattle

Empirical Example: Ekaterina Jardim et. al (2018)

Policy Background

Table 1: Minimum Wage Schedule in Seattle under the Seattle Minimum Wage Ordinance

Effective Date	Large Employers ^a		Small Employers	
	No benefits	With benefits ^b	No benefits or tips	Benefits or tips ^c
Before Seattle Ordinance				
January 1, 2015	\$9.47	\$9.47	\$9.47	\$9.47
After Ordinance				
April 1, 2015	\$11.00	\$11.00	\$11.00	\$10.00
January 1, 2016	\$13.00	\$12.50	\$12.00	\$10.50
January 1, 2017	\$15.00 ^d	\$13.50	\$13.00	\$11.00
January 1, 2018	\$15.45	\$15.00 ^e	\$14.00	\$11.50
January 1, 2019			\$15.00 ^f	\$12.00
January 1, 2020				\$13.50
January 1, 2021				\$15.00 ^g

Empirical Example: Ekaterina Jardim et. al (2018)

Data

- Administrative employment data from Washington State covering the period of 2005 through the third quarter of 2016
- Washington's Employment Security Department collects **quarterly payroll records** for all workers who received wages in Washington and are covered by Unemployment Insurance
- Employers are required to report actual hours worked
- They construct average hourly wage paid to each worker in each quarter by dividing total quarterly earnings by quarterly hours worked

Empirical Example: Ekaterina Jardim et. al (2018)

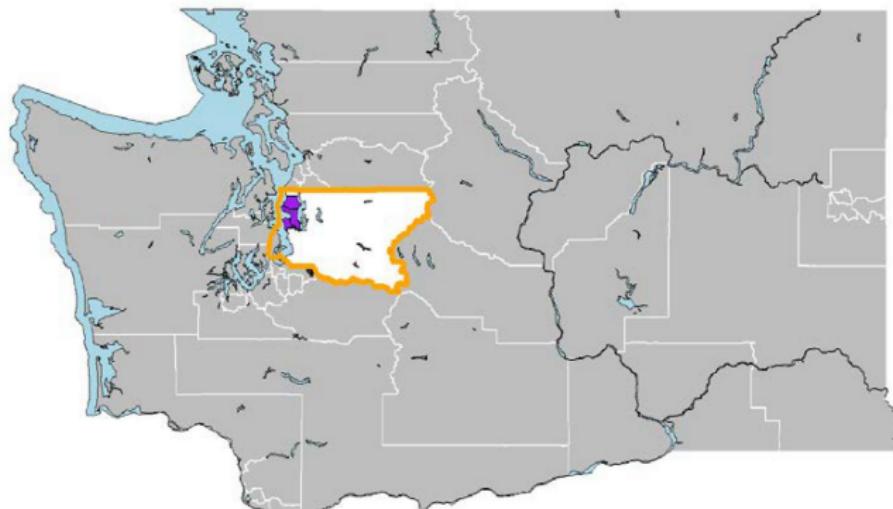
Research Design

- The unit of analysis is Public Use Microdata Areas (PUMAs) in Washington State
- A PUMA is a geographic unit defined by the U.S. Census Bureau with a population of approximately 100,000 people
- Seattle is composed of five PUMAs (treated regions)
- The remainder of Washington includes 40 PUMAs (control regions)

Empirical Example: Ekaterina Jardim et. al (2018)

Research Design

**Figure 4: Synthetic Control and Interactive Fixed Effects Regions
(Seattle and Public Use Microdata Areas Outside King County)**



Empirical Example: Ekaterina Jardim et. al (2018)

Identification Strategy: SC

- SC uses pre-policy observations to find an optimal set of (weighted) control regions to match pre-policy trend in the treated region

$$\min_{w_r} \sum_{t=-33}^0 (\Delta Y_{r=1,t} - \sum_{r=2}^R w_r \Delta Y_{rt})^2$$

- Subject to the constraints $\sum_r w_r = 1$ and $\forall r, w_r \geq 0$
- ΔY_{rt} : It represents year-over-year changes in each outcome in region r at period (year-quarter) t

- $\Delta Y_{rt} = \frac{Y_{rt}}{Y_{r,t-4}} - 1$

Empirical Example: Ekaterina Jardim et. al (2018)

Identification Strategy: SC

- Given a set of weights \hat{w}_r , the impact of minimum wage increase in period (year-quarter) q is estimated as follows:

$$\beta_q^{synth} = \Delta Y_{r=1,q} - \sum_{r=2}^R \hat{w}_r \Delta Y_{rt}$$

- Note that they focus on low-wage jobs (i.e., those earning under \$19 per hour)

Empirical Example: Ekaterina Jardim et. al (2018)

Results

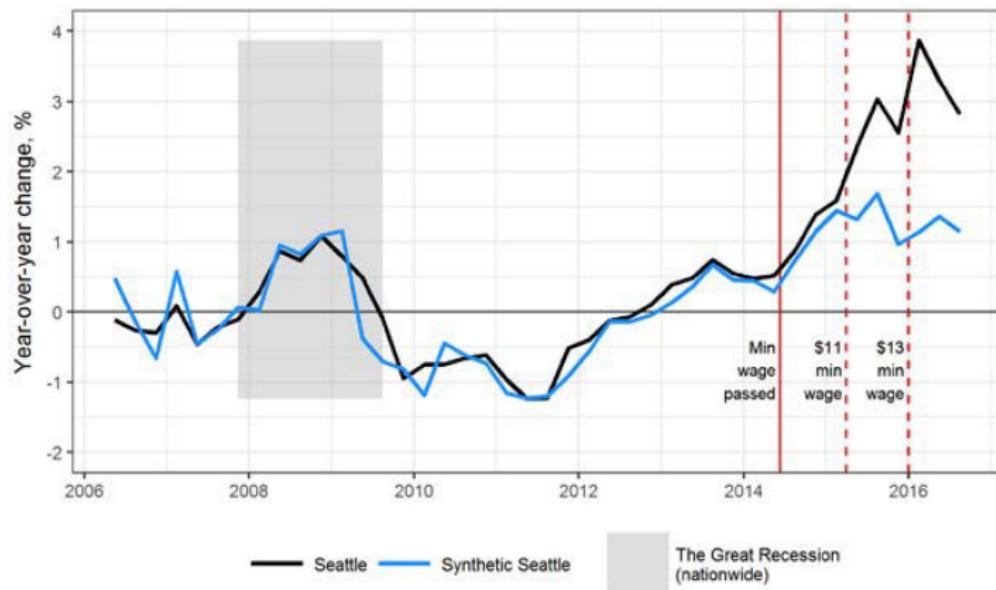
Table 5: Effect on Wages of Low-Wage Jobs

Quarter	Quarters After Passage / Enforcement	Synthetic Control	Interactive Fixed Effects
2014.3	1	0.002 [0.585]	0.005 [0.101]
2014.4	2	0.003 [0.465]	0.008*** [0.013]
2015.1	3	0.002 [0.598]	0.009*** [0.004]
2015.2	4/1	0.011** [0.029]	0.016*** [0.000]
2015.3	5/2	0.016*** [0.006]	0.022*** [0.000]
2015.4	6/3	0.019*** [0.000]	0.019*** [0.000]
2016.1	7/4	0.030*** [0.000]	0.032*** [0.000]
2016.2	8/5	0.031*** [0.000]	0.031*** [0.000]
2016.3	9/6	0.033*** [0.000]	0.034*** [0.000]
R2		0.781	
Pre-Policy RMSPE		0.003	
Obs.		1,890	1,890

Empirical Example: Ekaterina Jardim et. al (2018)

Results

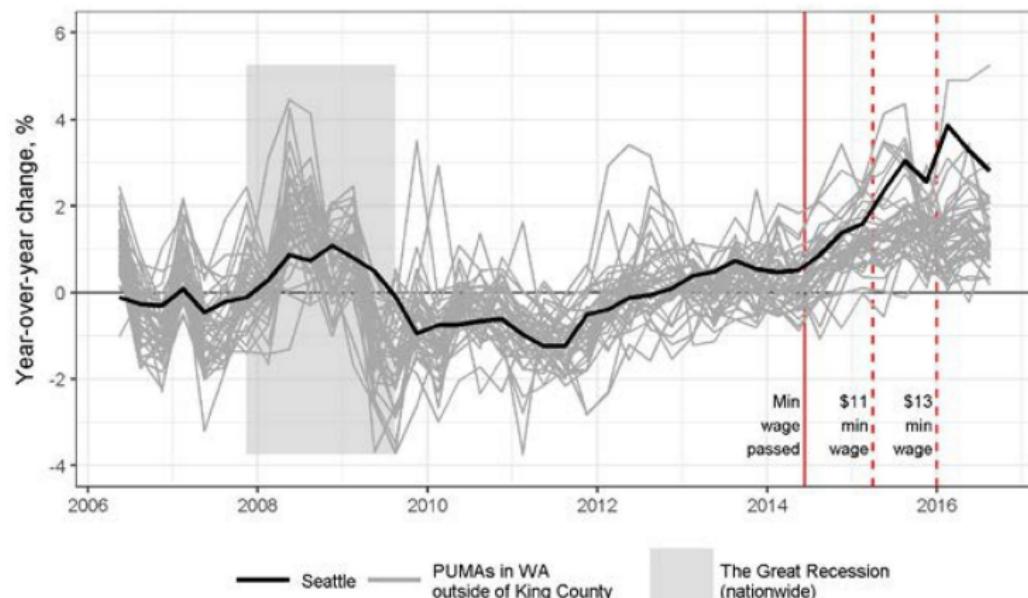
Panel A: Average Wage



Empirical Example: Ekaterina Jardim et. al (2018)

Results

Panel A: Average Wage



Empirical Example: Ekaterina Jardim et. al (2018)

Results

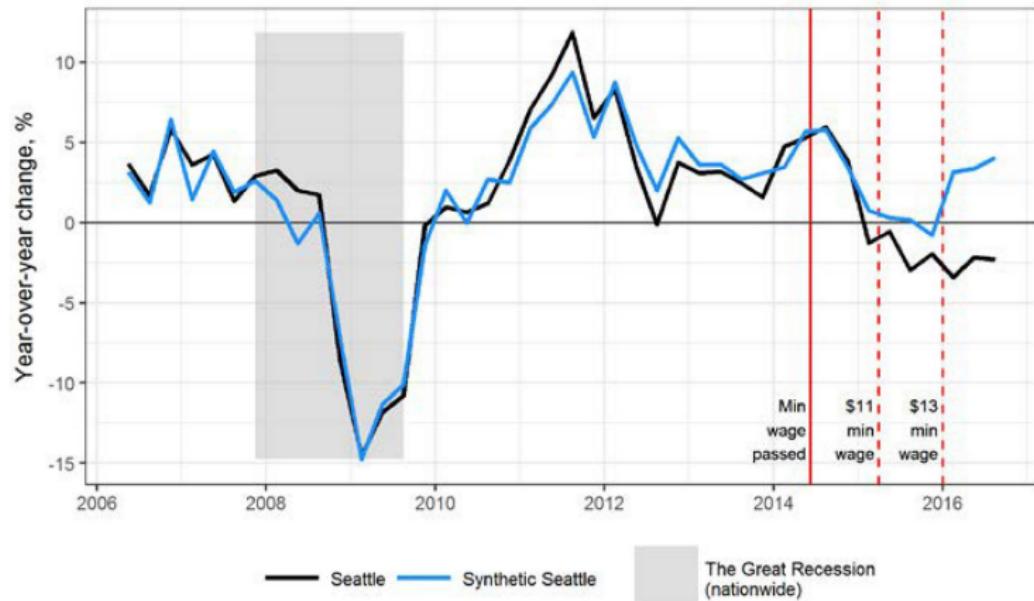
Table 6: Effect on Low-Wage Employment

Quarter	Quarters After Passage / Enforcement	Hours		Jobs	
		Synthetic Control	Interactive Fixed Effects	Synthetic Control	Interactive Fixed Effects
2014.3	1	0.002 [0.916]	0.005 [0.766]	0.002 [0.924]	-0.003 [0.842]
2014.4	2	0.006 [0.713]	0.000 [0.975]	-0.002 [0.892]	-0.014 [0.357]
2015.1	3	-0.018 [0.336]	-0.015 [0.349]	0.007 [0.659]	-0.005 [0.724]
2015.2	4/1	-0.006 [0.756]	-0.008 [0.594]	-0.010 [0.549]	-0.024 [0.107]
2015.3	5/2	-0.027 [0.356]	-0.008 [0.715]	-0.011 [0.576]	-0.026 [0.223]
2015.4	6/3	-0.006 [0.894]	0.008 [0.735]	-0.033 [0.391]	-0.035 [0.109]
2016.1	7/4	-0.087*** [0.005]	-0.057*** [0.014]	-0.038 [0.293]	-0.032 [0.146]
2016.2	8/5	-0.066*** [0.022]	-0.046* [0.052]	-0.052* [0.076]	-0.071*** [0.001]
2016.3	9/6	-0.092* [0.051]	-0.064*** [0.023]	-0.072* [0.067]	-0.088*** [0.001]
R2		0.791		0.718	
Pre-Policy RMSPE		0.013		0.013	
Obs.		1,890	1,890	1,890	1,890

Empirical Example: Ekaterina Jardim et. al (2018)

Results

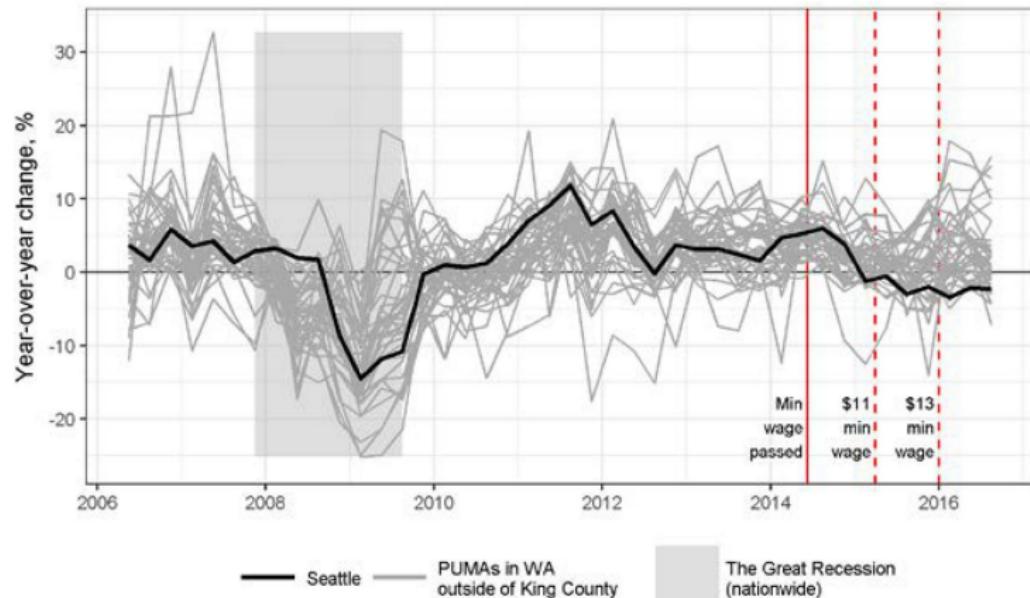
Panel B: Hours Worked



Empirical Example: Ekaterina Jardim et. al (2018)

Results

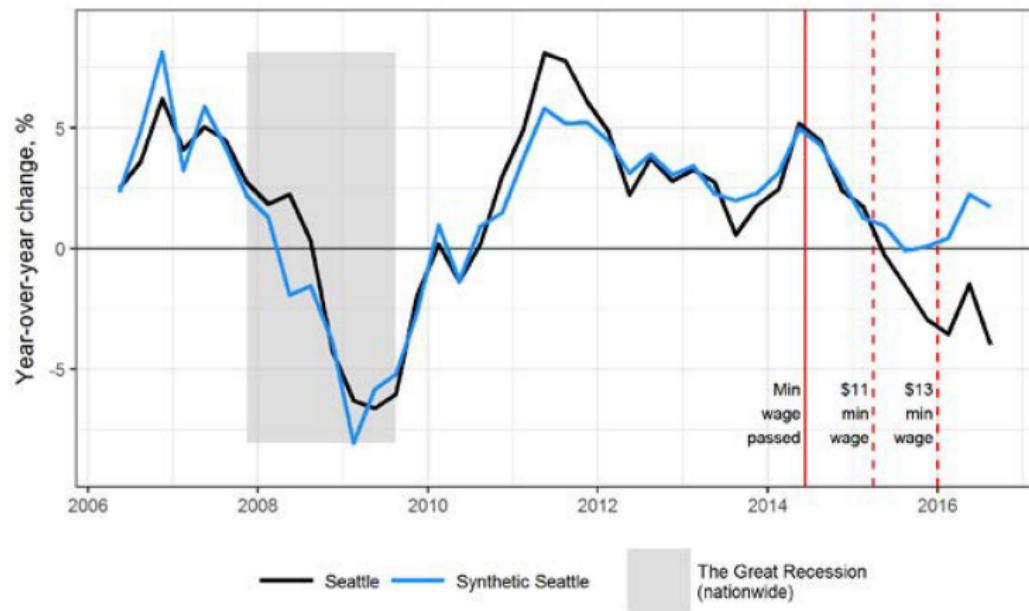
Panel B: Hours Worked



Empirical Example: Ekaterina Jardim et. al (2018)

Results

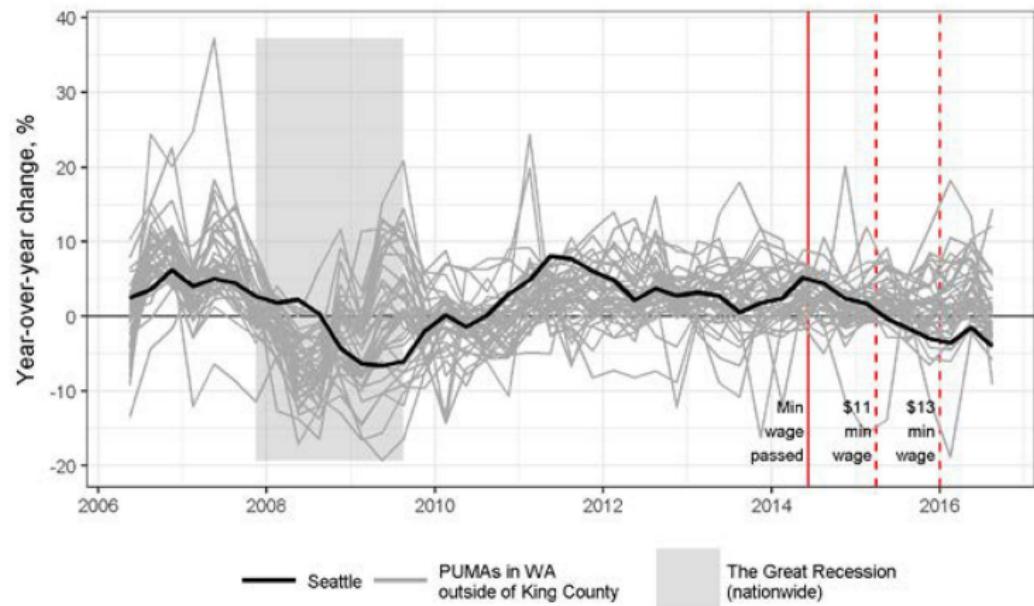
Panel C: Number of Jobs



Empirical Example: Ekaterina Jardim et. al (2018)

Results

Panel C: Number of Jobs



Empirical Example: Ekaterina Jardim et. al (2018)

Results

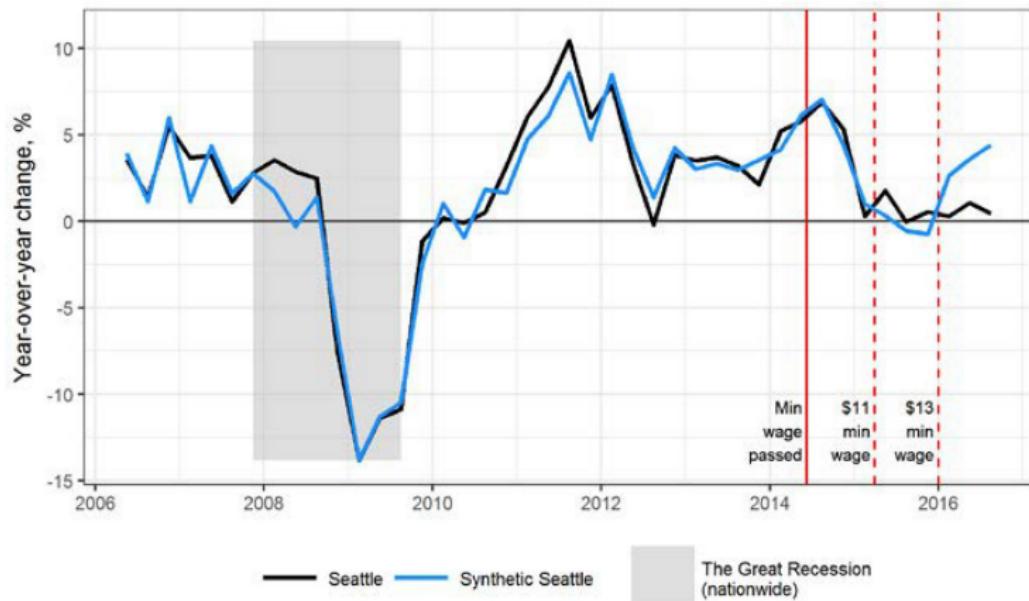
Table 7: Effect on Payroll for Low-Wage Jobs

Quarter	Quarters After Passage / Enforcement	Synthetic Control	Interactive Fixed Effects
2014.3	1	-0.001 [0.946]	0.014 [0.301]
2014.4	2	0.012 [0.479]	0.012 [0.404]
2015.1	3	-0.004 [0.836]	-0.006 [0.698]
2015.2	4/1	0.017 [0.399]	0.01 [0.486]
2015.3	5/2	0.006 [0.847]	0.015 [0.478]
2015.4	6/3	0.025 [0.614]	0.023 [0.286]
2016.1	7/4	-0.032 [0.416]	-0.035 [0.149]
2016.2	8/5	-0.013 [0.739]	-0.024 [0.352]
2016.3	9/6	-0.037 [0.519]	-0.039 [0.176]
R2		0.825	
Pre-Policy RMSPE		0.012	
Obs.		1,890	1,890

Empirical Example: Ekaterina Jardim et. al (2018)

Results

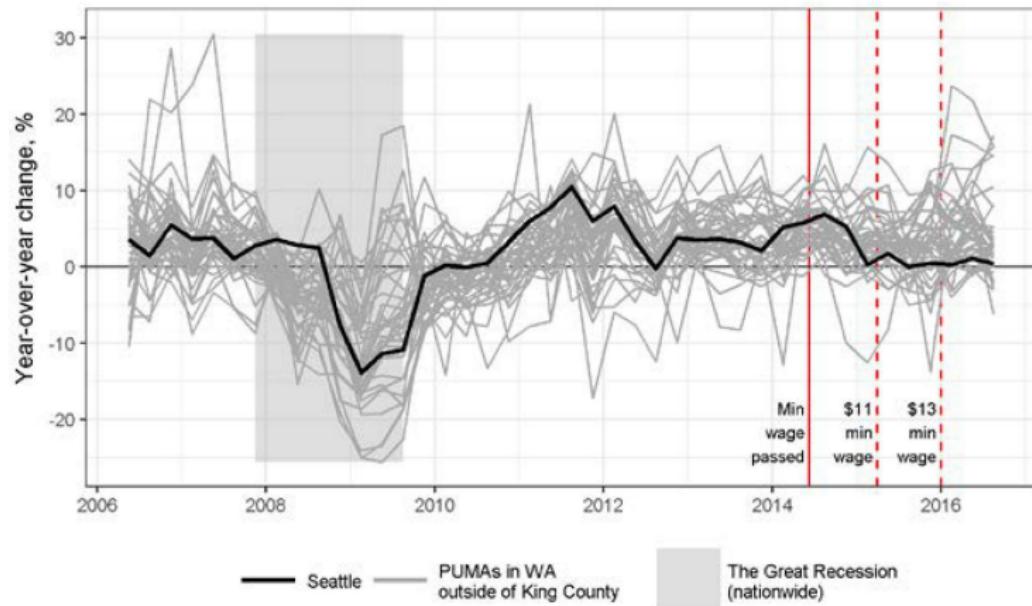
Panel D: Payroll



Empirical Example: Ekaterina Jardim et. al (2018)

Results

Panel D: Payroll



Empirical Example: Ekaterina Jardim et. al (2018)

Summary of Findings

- They conclude that the second wage increase to \$13 reduced hours worked in low-wage jobs by 6.9%
- Hourly wages in such jobs increased by 3.2%
- Importantly, the **lost income associated with the hours reductions exceeds the gain associated with the net wage increase**
 - The average low-wage employee was paid \$1,900 per month
 - The reduction in hours would cost the average employee \$130 per month
 - While the wage increase would recoup only \$56 of this loss
 - Leaving a net loss of \$74 per month, which is sizable for a low-wage worker