

# **Job Search and the Threat of Unemployment Benefit Sanctions**

Thomas Walsh  
European University Institute

Crossley Group

February 7, 2023

## Unemployment Insurance:

- smoothes consumption while searching
- insurance-incentive tradeoffs

## Sanctions policy tool to blunt the tradeoff:

- partial or complete stop of transfers for (e.g.) low search effort
- smoothing with less moral hazard
- jobseekers create worse matches ("market insurance")

## Channels:

- direct: punishment via budget constraint
- **indirect: threat, deterrent via expectations**
  - much larger group!
  - e.g: if 10% ever sanctioned, 9X larger

## **Policymakers tend to “toughen up” the UI regime after recessions** (GFC, Covid)

- “back to work” political rhetoric,
- fiscal budget / austerity: insurance is nice, but we can’t pay for everyone
- UK reform in 2012 (*other examples: UK, France, Germany 2022*)

① Does sanction threat change search behaviour?

② To what extent does sanction threat create worse matches?

⇒ To answer these questions empirical strategy uses **UK Sanction Policy Reform in 2012**

- exploit **differential responses** across districts in sanctioning rate
- spatial heterogeneity in sanction response lends itself to **Difference-in-Differences** design

# Snapshot of Results

Define: sanction threat as **sanctions-per-UI claimant** in the jobseeker's local area

- ① **Stylised Fact:** sanction threat correlates strongly with earnings losses in the medium term

Define: sanction threat as **sanctions-per-UI claimant** in the jobseeker's local area

- ① **Stylised Fact:** sanction threat correlates strongly with earnings losses in the medium term
- ② **Sanction threat increases exit speed from unemployment**
  - reform increases hazard rate  $\sim 20$  percent (100% direct effect  $\times 0.05-0.1$ )
  - a 1ppt increase in sanction threat raises exit hazard by 0.5 ppts (baseline: 4.5)

Define: sanction threat as **sanctions-per-UI claimant** in the jobseeker's local area

- ① **Stylised Fact:** sanction threat correlates strongly with earnings losses in the medium term
- ② **Sanction threat increases exit speed from unemployment**
  - reform increases hazard rate  $\sim 20$  percent (100% direct effect  $\times 0.05-0.1$ )
  - a 1ppt increase in sanction threat raises exit hazard by 0.5 ppts (baseline: 4.5)
- ③ **treated districts experience more spells of unemployment**
  - $\sim 10$  percent increase in cumulative spells

Define: sanction threat as **sanctions-per-UI claimant** in the jobseeker's local area

- ① **Stylised Fact:** sanction threat correlates strongly with earnings losses in the medium term
- ② **Sanction threat increases exit speed from unemployment**
  - reform increases hazard rate  $\sim 20$  percent (100% direct effect  $\times 0.05-0.1$ )
  - a 1ppt increase in sanction threat raises exit hazard by 0.5 ppts (baseline: 4.5)
- ③ **treated districts experience more spells of unemployment**
  - $\sim 10$  percent increase in cumulative spells
- ④ **treated districts have worse reemployment durations**
  - $\sim 5-10$  percent less likely to reach 12/24/36 months continuously employed



# Existing Literature and This Work

- **A UI system matters for eqm job characteristics, match quality/suitability to skills:**  
Acemoglu (2001); Acemoglu and Shimer (1999, 2000); and Marimon and Zilibotti (1999).
- **...and UI duration for reemployment wages**  
Nekoei, Weber (2017) Card, Chetty, Weber (2007)
- **Direct Sanction Effect on Exit from unemployment** in most studies (typically northern EU)  
Abbring & Van Den Berg\* (2003) Boockmann, Thomsen, Walter (2014) Arni, Lalive, van Ours (2013), van der Klaauw & van Ours (2011) Boone, Sadrieh, van Ours (2009), Lalive, Zweimueller, van Ours (2005) Lombardi (2019)\*
- **...and inactivity** Petrongolo (2009)
- **some evidence local average sanction rates** also matter Lombardi, 2019

# Existing Literature and This Work

- **A UI system matters for eqm job characteristics, match quality/suitability to skills:**  
Acemoglu (2001); Acemoglu and Shimer (1999, 2000); and Marimon and Zilibotti (1999).
- **...and UI duration for reemployment wages**  
Nekoei, Weber (2017) Card, Chetty, Weber (2007)
- **Direct Sanction Effect on Exit from unemployment** in most studies (typically northern EU)  
Abbring & Van Den Berg\* (2003) Boockmann, Thomsen, Walter (2014) Arni, Lalive, van Ours (2013), van der Klaauw & van Ours (2011) Boone, Sadrieh, van Ours (2009), Lalive, Zweimueller, van Ours (2005) Lombardi (2019)\*
- **...and inactivity** Petrongolo (2009)
- **some evidence local average sanction rates** also matter Lombardi, 2019

## To differentiate my work:

- focus on indirect effect
- approach identification from a new longitudinal angle (DID vs "timing of events")

# Policy Reform Details

## **Unemployment Insurance in UK**

- search not duration or contribution contingent

## **Possible Reasons for a sanction:**

- Failure to attend advisor meeting / work program
- Unavailable to work
- Ineligible search effort
- Refusing, voluntarily leaving work
- Dismissal for misconduct

# UK Sanctions Regime and Reform

## Features of UI/sanctions:

- sanction = UI payments **stopped** for N weeks
- About 70 GBP/week (80 EUR) , flat over time in real terms.
- Replacement rates low with/without other transfers. Adjusted for age (25) and couples

## Effects of “Toughening Up” Reforms:

- large increases in sanctions-per-unemployed **estimated +40%** of baseline

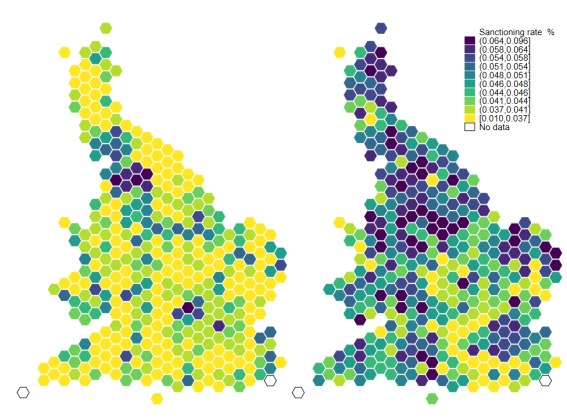
## Sources of Variation:

- degree of job centre discretion/autonomy
- use of sanction/exit targets

*National Audit Office:* “The NAO concludes it is likely that management focus and local work coach discretion have had a substantial influence on whether or not people are sanctioned (...) heterogeneity [in sanction rates across areas] not fully explained by jobseeker characteristics”

# Reform Effect on Sanction Intensity

Figure 1: Sanctioning Rates (% , pre vs post)



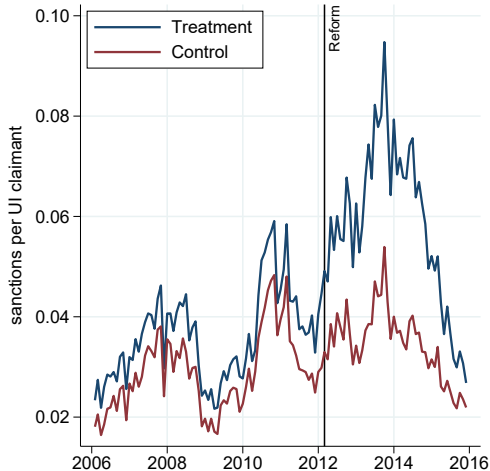
$$S_{gt} = \frac{\text{number of sanctions issued}_{gt}}{\text{number of UI claimants}_{gt}}$$

Who ends up tough, post-reform?

- very similar levels in  $t$
- Treated: upper 25% in  $t + 1$
- Control: lower 25% in  $t + 1$
- very correlated with changes  $(t, t + 1)$

# Identifying Variation

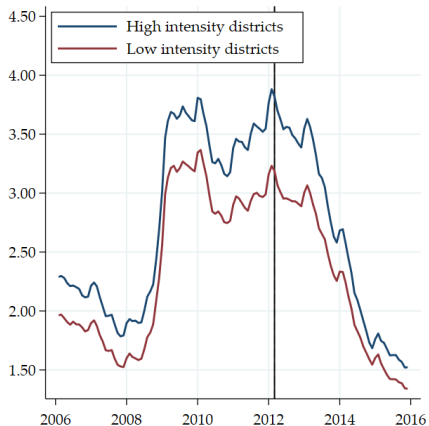
Figure 2: sanctioning rate (%)



- Treatment defined in data-driven way
- *Threat*: local shocks drive policy actions
- examine other variables to look for shocks

Figure 3: Equilibrium Labour Market Conditions

(a) Unemp-Population Ratio (%)



(b) Weekly Earnings

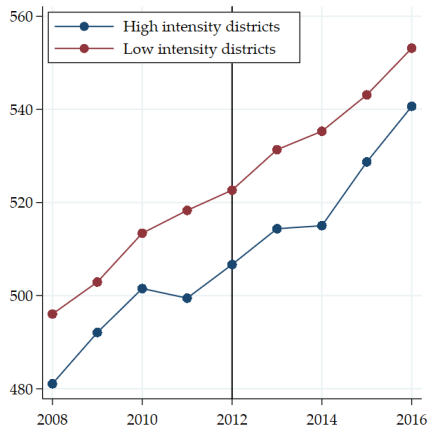
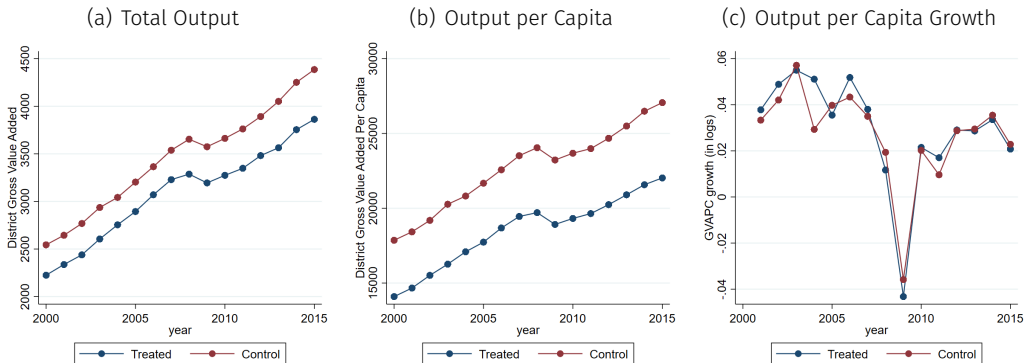


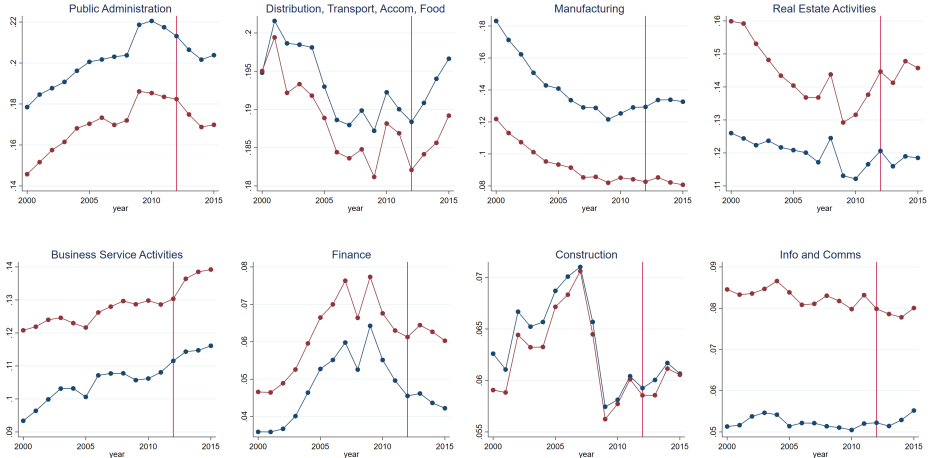


Figure 4: District-level Output (Real GVA)



\*excludes Westminster and City of London due to high business concentration

Figure 5: District-Industry Output Shares ( $GVA_{ind,dist,year}/GVA_{dist,year}$ )



## Local idiosyncratic shocks (reverse causality)

- Pre-trends cannot guarantee local shocks in 2012 didn't drive policy response
- Test economic conditions for diverging outcomes

## Migration driven by policy

- **mixing**: high to low unemployment, T-C gap closes, biased towards zero
- **polarising**: high to higher S, low to lower S, estimates biased away from zero

## Spillovers across space

- weekly jobseeker-advisor meetings, will be updated on current, local sanctioning rate
- T-C gap would close, estimates biased towards zero (again, if mixing spillovers)
- high degree of clustering of T/C poses a problem

## 1. Working Life Histories extracted from UKHLS ("*Understanding Society*")

- monthly history of labour market activity of around 40,000
- annual: income, hours, occupation, transfers, commute time, search effort
- (+) can potentially see items not tracked in admin data
- (-) loss of precision of exact month of transitions

### Sample:

- 60,000 obs,
- 2009m1-2015m12
- restricted to ages 18-64
- median duration 9m, mean 12m, 68% below 1y

## 2. District-level variables and National macro variables sanction information etc.

→ Matched assuming no district changes between waves, can't see any very short-term moves.

## Activity state by month:

- *"Which best describes your current situation?"*
  - Self-employed
  - In paid employment
  - **Unemployed**
  - Retired
  - On maternity leave
  - Caring for family/home
  - FT student
  - LT sick, disabled
  - Gov training scheme
  - Unpaid worker in family business
  - Apprenticeship
  - Something else
- transitions map well to "event studies" and national statistics (LFS)

► transitions

# Displacement Earnings Losses and Sanction Threat

# Earning losses due to job displacement

## Displacement event study regression:

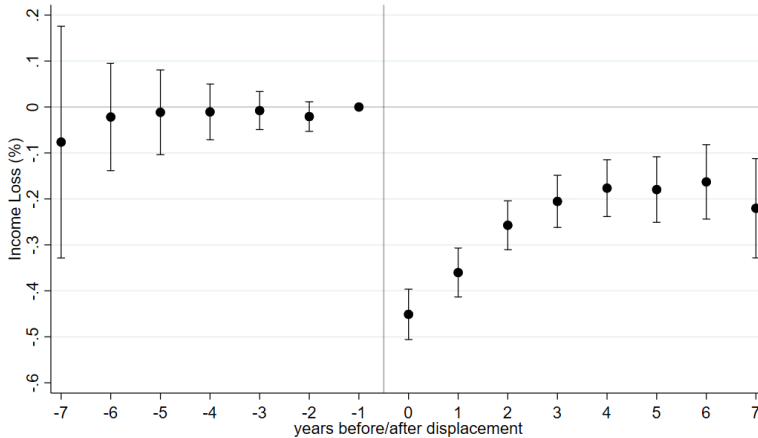
- Staggered Diff-in-Diff (job loss in different years)
- stacking estimator of Cengiz et al (QJE,2019), combines many  $2 \times 2$  diff-in-diffs
- makes clean comparisons of displaced vs not-yet-displaced / i.e. is *stagger-robust*

$$y_{ict}(r) - y_{ict}(-1) = \lambda_t + \sum_{r=-7}^7 \alpha_r \mathbb{1}_{\{r\}} + \sum_{r=-7}^7 \beta_r (\mathbb{1}_{\{r\}} \times \text{displaced in year } c_{ic}) + \varepsilon_{it} \quad (1)$$

- normalised to  $r = -1$  in relative event time

► stacking regression

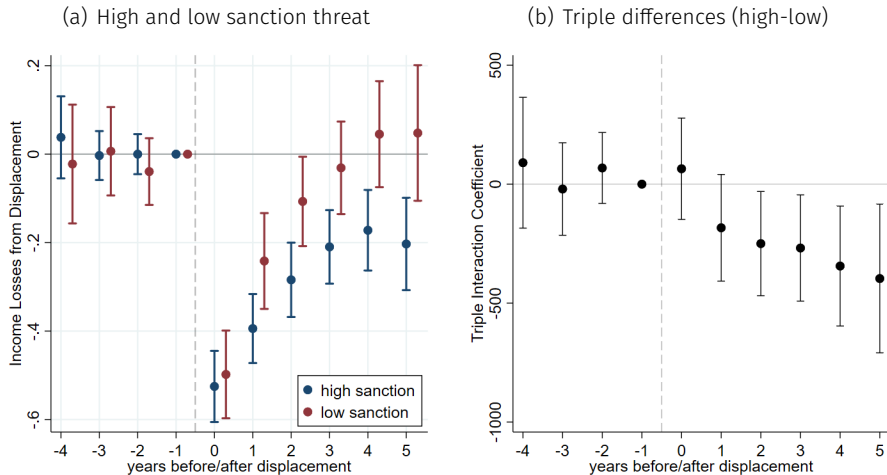
Figure 6: Earnings Losses from Job Loss



Sample: Ever-displaced only. Treated: lose job in year  $t$ , control: not-yet-treated by  $t$ .  
Excludes zero earnings. Including zeroes leads to approx -40pct.

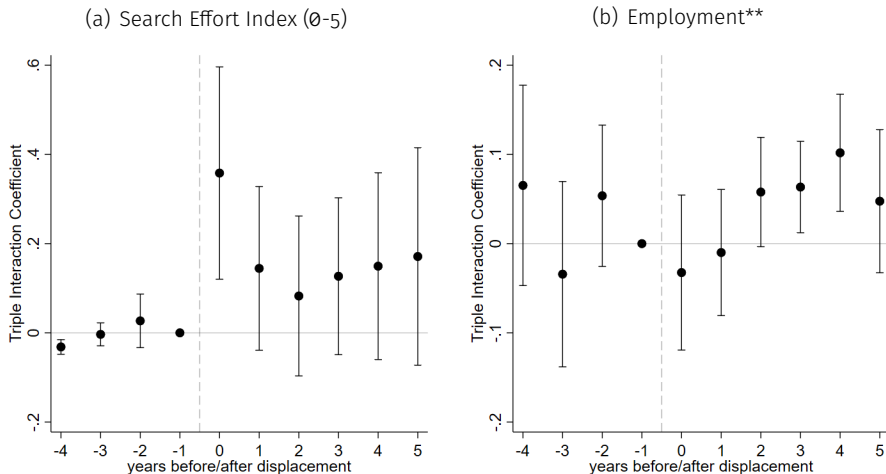


Figure 7: Earnings Losses by high/low sanction threat in early unemployment



High sanction: average sanction rate in first 3 months of spell above/below average

Figure 8: Triple-Differences Estimates comparing displacements with high vs low sanctioning



less conservative sample restriction: employed in  $r = -1$  only. \*\*Employed at time of survey\*\*

# Causal Estimates

## **Difference-in-differences design**

- use changes in controls to infer changes that would have happened in treatment group, absent treatment

## **Identifying variation in the data**

- exploit heterogeneity in intensity responses to common reform
- T,C selected in a data-driven way

# Identifying assumptions

## A1. No Spillovers

- neither *across space* nor *through time*
- outcome depends only on *own contemporaneous treatment status*

## A2. Common Trends.

- Absent treatment, Treated and Controls would have followed the same **changes** in outcomes
- DiD can tolerate a degree of endogeneous treatment
- can't handle contemporaneous reforms or asymmetric shocks

**(A1+A2+panel data):** ATT/average treatment effect *on the treated* is identified

► assumptions

# Estimating Equation

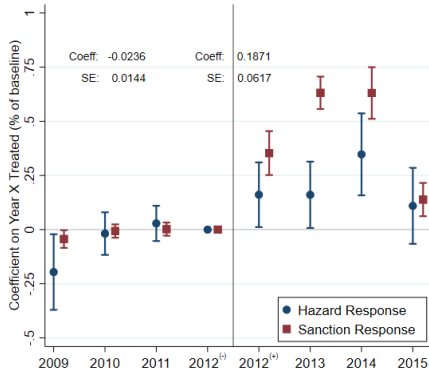
Canonical 2x2 simultaneous-reform Difference-in-differences:

$$\mathbb{1}\{\textit{exit}\}_{igt} = \underbrace{\lambda_t + \gamma_g + \theta_{T(i,t)}}_{\text{common trends plus duration}} + \underbrace{\sum_{\ell=-4}^4 \beta_{\ell} \times (\mathbb{1}\{t = \ell\} \cdot \textit{Treated District}_g)}_{\text{placebos/ATTs}} + u_{igt} \quad (2)$$

- $\theta_{T(i,t)}$  duration-of-spell effects
- two-stage estimation.  $(\lambda, \gamma, \theta)$  using  $(D_{gt} = 0)$  obs (Gardner, 2022)
- if treatment affects exit, it necessarily changes duration-of-spell

Figure 9: Difference-in-Differences Estimates

(a) Exit rate, total (% of baseline rate)



(b) Exit rate, into employment (% of baseline)

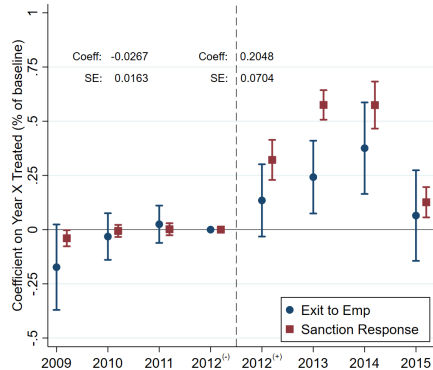
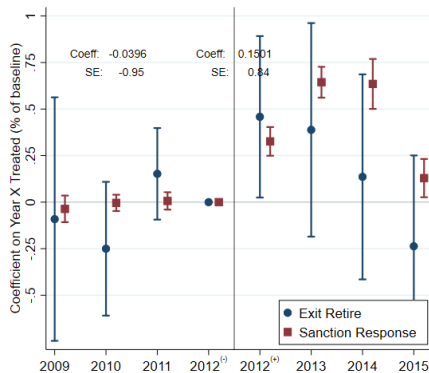


Figure 10: Exit to retirement

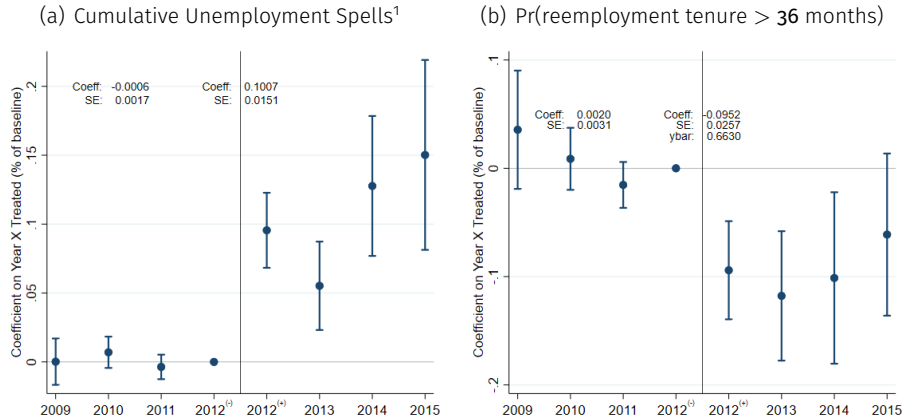
(a) Exit to retirement (% of baseline rate)



Post-reform spike one-and-done effect. Very low precision.



Figure 11: Reemployment Stability



<sup>1</sup>since start of sample

Table 1: Regression Results: ATT estimates

	Exit rate			Unemp.	Re-employ. duration			Sanction
	total	employed	retired	$N_u$	>12m	>24m	>36m	
$\beta$ (ppts)	0.00860*** (3.03)	0.00796*** (2.91)	0.00455 (0.84)	0.112*** (6.65)	-0.0396*** (-2.85)	-0.0410** (-2.45)	-0.0631*** (-3.70)	0.0170*** (14.92)
$\beta$ (%)	0.191*** (3.03)	0.205*** (2.91)	0.150 (0.84)	0.101*** (6.65)	-0.0504*** (-2.85)	-0.0583** (-2.45)	-0.0952*** (-3.70)	0.418*** (14.92)
NT	59070	59070	12696	59070	59070	59070	59070	58672

## Main takeaways:

- Sanctioning policy acts on a **wide set of job-seekers**, not just the directly punished.
- Effects go **beyond exit rates**.
- **fast exits, less stable jobs with more unemployment risk**  
⇒ optimal UI design with endogenous unemployment risk
- **implies a intensive-extensive margin tradeoff: duration of spell vs number of spells**  
(⇒ welfare implications)

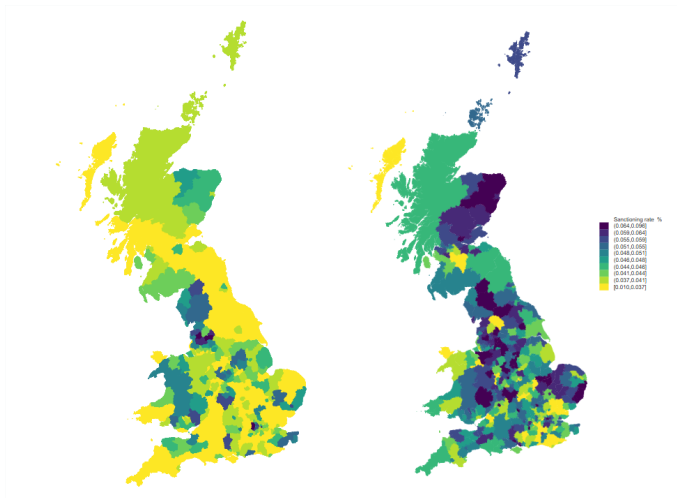
## Main takeaways:

- Sanctioning policy acts on a **wide set of job-seekers**, not just the directly punished.
- Effects go **beyond exit rates**.
- **fast exits, less stable jobs with more unemployment risk**  
⇒ optimal UI design with endogenous unemployment risk
- **implies a intensive-extensive margin tradeoff: duration of spell vs number of spells**  
(⇒ welfare implications)

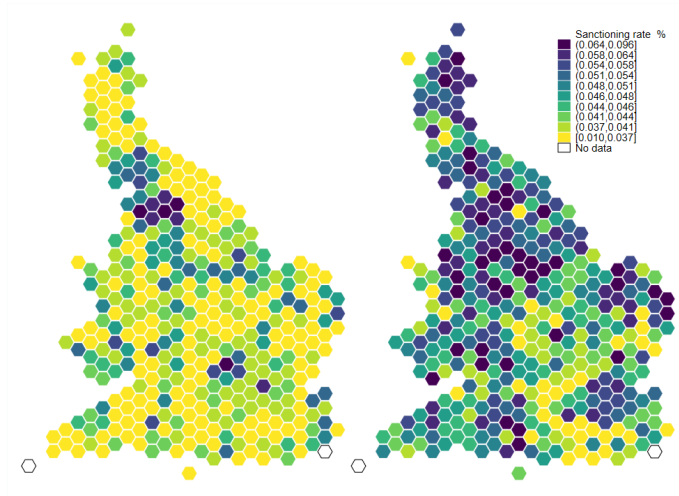
*Thanks! [thomas.walsh@eui.eu](mailto:thomas.walsh@eui.eu)*

# Appendix

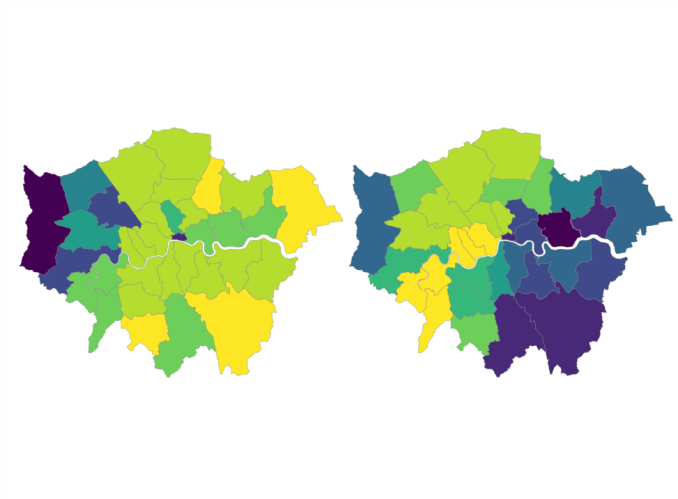
# Mapping Sanction Rates, 2010/12 vs. 2012/14



# Mapping Sanction Rates, 2010/12 vs. 2012/14

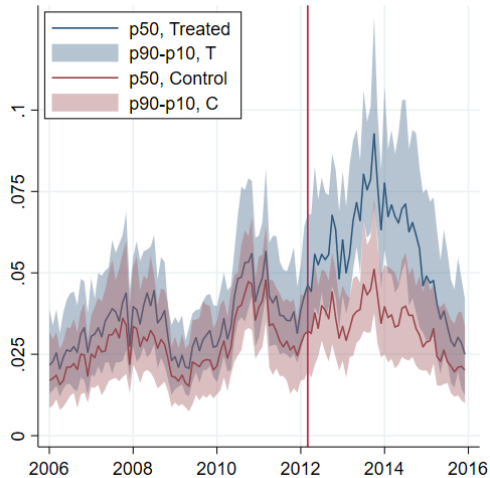


# Mapping Sanction Rates, 2010/12 vs. 2012/14; London





# Distribution on Sanction Rates within Group



# Intensive Margin of Sanction Reform

Infraction Level	Example Reasons	Sanction in weeks	
		<i>Before</i>	<i>After</i>
Lower	Failure to attend advisor meeting Failure to attend work program	1	4,13
Intermediate	Unavailable to work Ineligible search effort	0	4, 13
Higher	Refusing, voluntarily leaving work Dismissal for misconduct	1-26	4, 26, 156

Table 2: **Structure of Sanctions**

# Identifying assumptions

**Potential outcomes:**  $Y_{it}(\mathbf{D})$

**A1. No Spillovers** neither across space nor through time

$$Y_{it}(\mathbf{D}) = Y_{it}(D_{gt} \in \{\emptyset, 1\})$$

**A2. Common Trends.**

Absent treatment, Treated and Controls would have followed the same **changes** in outcomes

$$E[Y_{it+1}(\emptyset) - Y_{it}(\emptyset) | D_i = 1] = E[Y_{it+1}(\emptyset) - Y_{it}(\emptyset) | D_i = \emptyset]$$

**(A1+A2+panel data):** ATT/average treatment effect *on the treated* is identified

$$E[Y_{it+1}(1) - Y_{it+1}(\emptyset) | D_i = 1]$$

# Stacking Estimator

- Suppose treatment in each  $t$  is a different intervention (losing a job in 1998  $\neq$  in 2002)
- combine many small  $2 \times 2$ s

$$D_{98}(g, t) = \begin{pmatrix} t & 97 & 98 & 99 & 00 & 01 & 02 \\ g_{97} & 1 & 1 & 1 & 1 & 1 & 1 \\ g_{98} & 0 & 1 & 1 & 1 & 1 & 1 \\ g_{99} & 0 & 0 & 1 & 1 & & \\ g_{00} & 0 & 0 & 0 & 1 & & \\ g_{01} & 0 & 0 & 0 & 0 & 1 & \\ g_{02} & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$D_{00}(g, t) = \begin{pmatrix} t & 97 & 98 & 99 & 00 & 01 & 02 \\ g_{97} & 1 & 1 & 1 & 1 & 1 & 1 \\ g_{98} & 0 & 1 & 1 & 1 & 1 & 1 \\ g_{99} & 0 & 0 & 1 & 1 & & \\ g_{00} & 0 & 0 & 0 & 1 & & \\ g_{01} & 0 & 0 & 0 & 0 & 1 & \\ g_{02} & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \quad (3)$$

- Can drop never-treated (NT)
- Top row doesn't have a control (always-treated, AT)
- Last row also left out (last-treated, LT)

# Aligned on Event Time

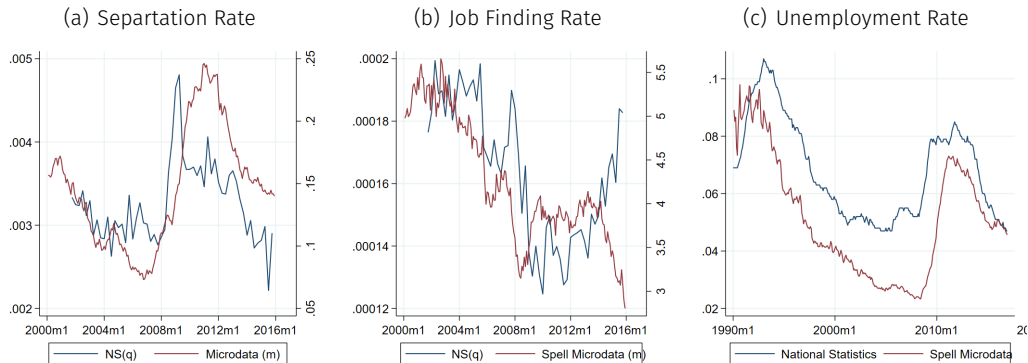
Aligning and averaging the 2x2DiDs

$$D(g, t) = \left( \begin{array}{c|ccccc|ccccc} r & -4 & -3 & -2 & -1 & = 0 & +1 & +2 & +3 & +4 \\ \hline g_{98} & & & & 0 & 1 & 1 & 1 & 1 & 1 \\ g_{99} & & & 0 & 0 & 1 & 1 & 1 & 1 & \times \\ g_{00} & & 0 & 0 & 0 & 1 & 1 & 1 & \times & \times \\ g_{01} & 0 & 0 & 0 & 0 & 1 & 1 & \times & \times & \times \end{array} \right) \quad (4)$$

- $R = 8$  relative periods from  $T = 6$
- can see why event-studies have widening error bands

◀ back

Figure 12: Transition Rates



**1. “Job Search and the Threat of Unemployment Benefit Sanctions”**

job search and unemployment / job quality / policy effectiveness

**2. “Sectoral Volatility and the Investment Channel of Monetary Policy”**

dispersion of productivity shocks / policy effectiveness

**3. “Government Spending Multipliers in Firm-level Production Networks”**

distribution of procurement / policy effectiveness / budget efficiency

**4. “Making the Cut: Close Elections and Local Welfare Policies”**

political origins of policy asymmetry in UK

---

Governance of Institutions and Systems

- Institutions, Incentives and Welfare
- Public Finance
- Political Economy

(Social Determinants of Health)

- information, taxes and incentives)