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# Discussion of “Non-Disclosure Agreements and the Market for Talent”

Alexander Ljungqvist



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  - Policy bundling
  - Mechanism credibility
  - Equilibrium interpretation



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**Bottom line:** Strong paper, careful design, interesting setting.



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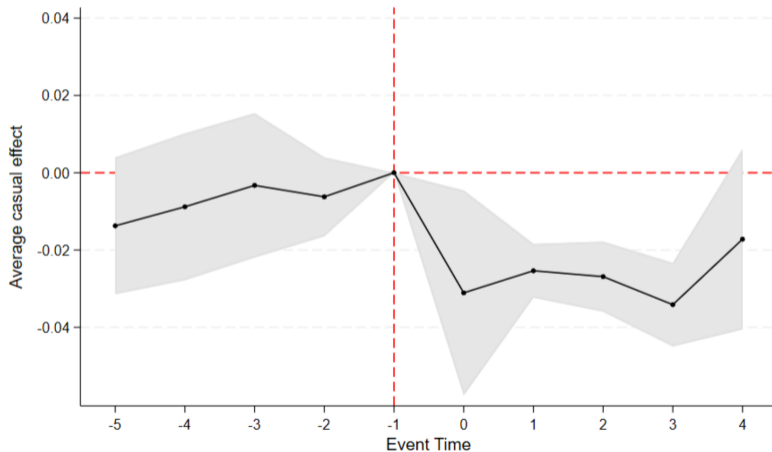
## Empirical setting:

- State laws restricting NDAs (2018–2020)

## Main question:

- Do these laws improve firms' ability to attract and retain workers?

Figure 2: Narrowing NDAs' Dynamic Effect on Job Vacancy Duration: Stacked DID Method





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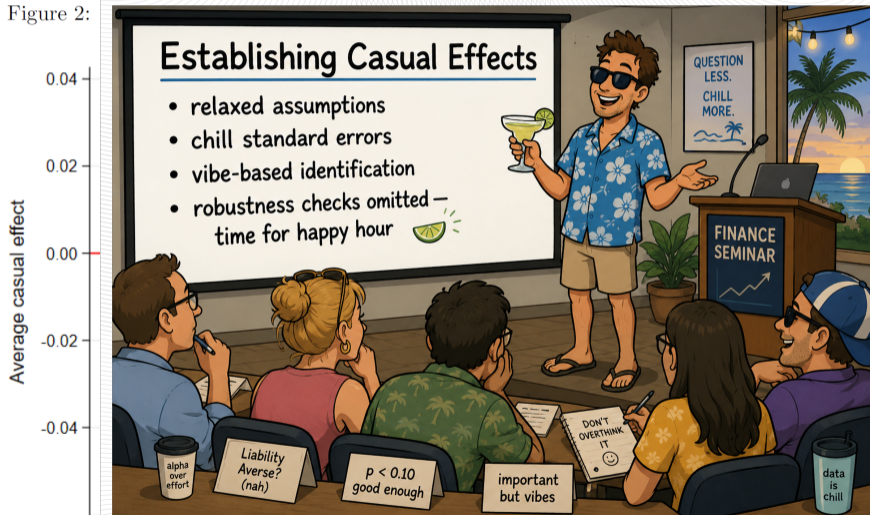
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- Relaxed assumptions
- Chill standard errors
- Vibe-based identification
- Robustness checks omitted – time for happy hour

Figure 2:







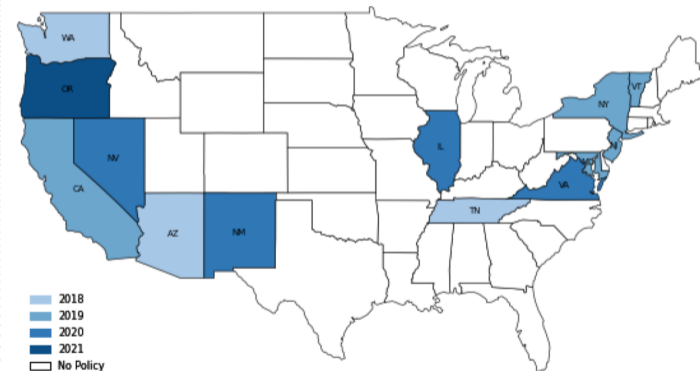
- Vacancy duration decreases → faster hiring
- Workplace conditions improve (injuries, violations)
- Job search activity increases
- Employee retention improves
- Firm performance (ROA) increases

## **Interpretation:**

- Transparency improves matching and incentives



Figure B1: Staggered Adoption of State Policies Narrowing NDAs



Treat State	Treated Year
Arizona	2018
Tennessee	2018
Washington	2018
New York	2019
Vermont	2019
Maryland	2019
California	2019
New Jersey	2019
Nevada	2020
Virginia	2020
Illinois	2020
New Mexico	2020
Oregon	2021



## Difference-in-differences design

- Treatment: state-level NDA restrictions
- Outcome: job vacancy duration

**Key identifying variation:** Within-firm, cross-state comparisons

## Controls:

- Firm-state-occupation FE
- Firm-year-occupation FE

## Additional designs:

- Event study (no pre-trends)
- Border-county comparison
- Stacked DiD



## **Related literatures:**

- Employee voice / disclosure (Glassdoor, whistleblowing)
- Labor market matching and information frictions
- Non-wage job characteristics

## **Key contribution:**

- Clean policy shock (NDA laws)
- Links transparency → workplace conditions → recruiting outcomes → firm performance

## **Takeaway:**

- A nice “full-chain” empirical paper



- Rich fixed effects (firm-state-occupation, firm-year-occupation)
- Large-scale, granular job posting data
- Multiple robustness checks:
  - Border-county design
  - Stacked DiD
- Mechanism evidence (workplace conditions, search activity)

## Overall:

- Aligned with current empirical standards



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**Recommendation: Make the stacked DiD the baseline**

## Implementation question

“Controls = neighboring states” (except CA)

Do you ensure that **controls** are only used if they are **not-yet-treated** throughout the event window?



**Key principle: Controls must be not yet treated at each time**

Year	2016	2017	2018	2019	2020	2021
State A (treated 2018)	Pre	Pre	Post	Post	Post	Post
State B (treated 2020)	Pre	Pre	Pre	Pre	Post	Post

**Using State B as control for A:**

- 2016–2019: **Valid (not yet treated)**
- 2020–2021: **Invalid (already treated)**

**Takeaway:**

- Controls must drop out once they are treated
- Otherwise: treated units serve as controls → bias



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- Especially problematic when many observations are 0 or 1 → large mass of zero in violations, injuries?
- Hard to interpret economically: not a semi-elasticity
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**Best practice (Santos da Silva & Tenreyro REStat 2006):**

- “The log of gravity”: Respect the count nature of the data
- Estimate Poisson Pseudo-Maximum Likelihood (PPML – can handle high-dimensional FE)
- Or: negative binomial (may not work here given the FE structure)



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**Identifying assumption:**

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- Worker mobility across states
- Multi-state firms changing policies nationally
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**Suggestion:**

- Explore heterogeneity: single-state firms, geographically isolated markets



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### **Possible extensions:**

- Control for other contemporaneous state labor policies (anti-harassment legislation, arbitration restrictions, pay-transparency laws, broader worker-protection reforms...)
- Show results robust to excluding states with especially broad MeToo-related reform packages



## Mechanism is fundamentally informational:

- NDA → more credible disclosure → better worker info → firms and workers respond + matching improvements

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## Suggestions:

- Directly measure disclosure at firm level. Show disclosure increases, **then** workplace conditions improve (i.e., establish timing)
- Exploit heterogeneity in ex ante opacity and/or in pre-treatment conditions



## **Outcome: vacancy duration**

- Reflects both supply and demand

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## **Remaining questions:**

- General equilibrium labor reallocation?
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## **Suggestions:**

- Explore geographic heterogeneity
- Discuss equilibrium more explicitly



- Transparency matters for labor markets
- Information frictions affect both:
  - matching
  - firm incentives
- Policy can improve both worker and firm outcomes

## **Contribution:**

- Clean policy setting + strong empirical design



## Strengths:

- Interesting question
- Strong data and design
- Rich set of outcomes

## Suggestions:

- Address spillovers (SUTVA)
- Further isolate policy variation
- Strengthen mechanism evidence

## Overall:

- A compelling and important paper