

Work from Home Before and After the Covid-19 Outbreak

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- Novel dataset tracking US labor market and WFH around COVID-19
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 - ▶ Micro data publicly available at openICPSR

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 - ▶ Measured WFH: workdays without a commute
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- Aggregate changes in WFH
 - ▶ Pre-covid: 15% workdays at home → May '20: 40% → June '21: 29%
- Who drove the increase in WFH during the pandemic?
 - ▶ Job-stayers, pre-pandemic daily commuters, in a few skilled industries

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 - ▶ **Substitution**: Pandemic increased costs of working on-site
 - ▶ **Adoption**: Pandemic led to greater access to WFH work arrangements

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- Why is rise in WFH persistent?
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 - ▶ **Adoption**: Pandemic led to greater access to WFH work arrangements
- Quantitative model analysis
 - ▶ 3/4 of WFH increase in June 21 reflects adoption of new WFH arrangements
 - ▶ Full-time WFH share doubles to 14.6 percent in the longer term
 - ▶ One in 5 workdays WFH (up from 1 in 7 before the pandemic)
 - ▶ Consistent with WFH survey expectations and follow-up surveys

Outline

- ① The Real-Time Population Survey
- ② WFH Before and After the COVID-19 Outbreak
- ③ Model-Based Decomposition of Substitution and Adoption
- ④ Validation of Model Predictions

The Real-Time Population Survey (RPS)

- Novel monthly online survey spanning March 2020 - June 2021 (ages 18-64)
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- **Extra** information relative to CPS
 - ▶ Retrospective questions on February 2020 \implies **quasi-panel** data
 - ▶ New survey questions, including on **commuting behavior**

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- Do-it-yourself CPS: Bick and Blandin (RED 2022)

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How the RPS Measures Work from Home (WFH)

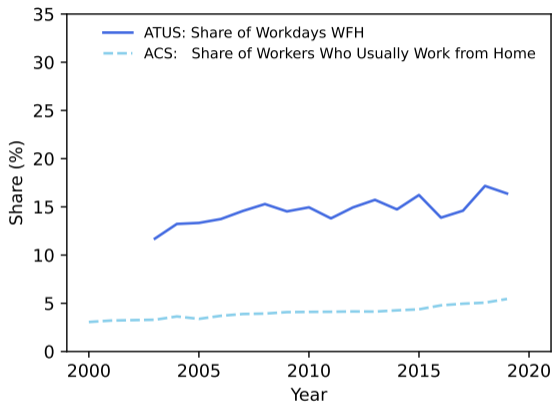
(intro)

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 1. *Last week, how many days did you **work** for this job?*
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 - ▶ **Commute-Only**: Days commuted = Days worked
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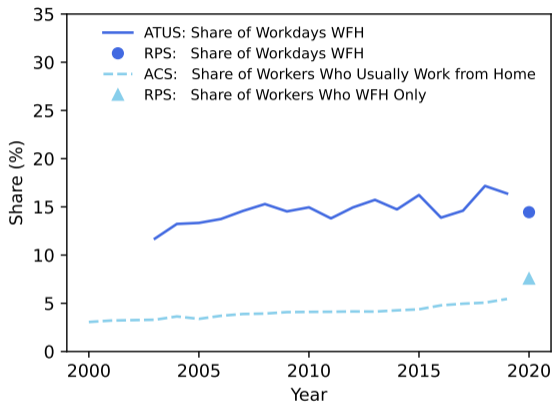
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- Note: WFH excludes days partly WFH, secondary jobs, home production

WFH Before COVID-19



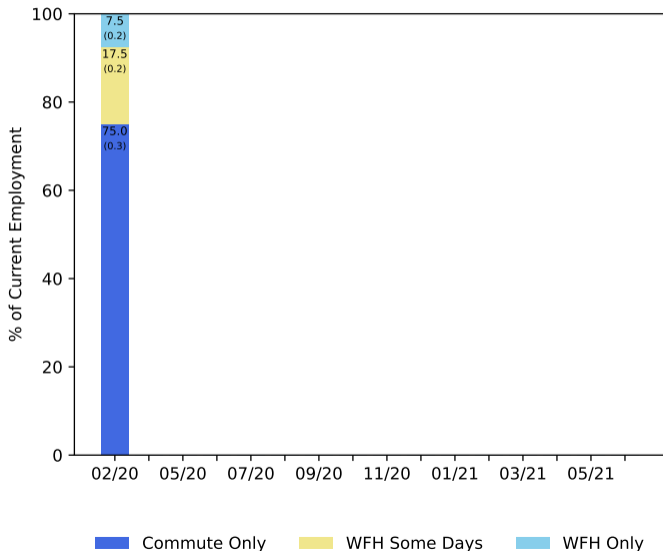
- Fairly slow increase in WFH over last two decades

WFH Before COVID-19

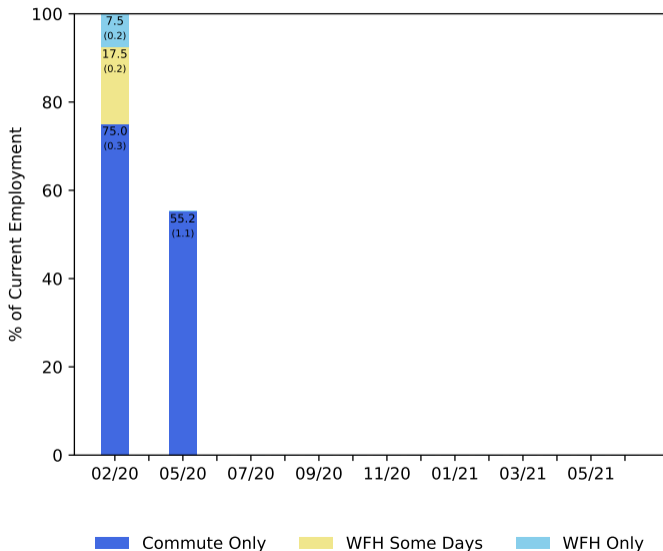


- Fairly slow increase in WFH over last two decades
- Feb. 2020 RPS aligns closely w/ 2019 ATUS, ACS

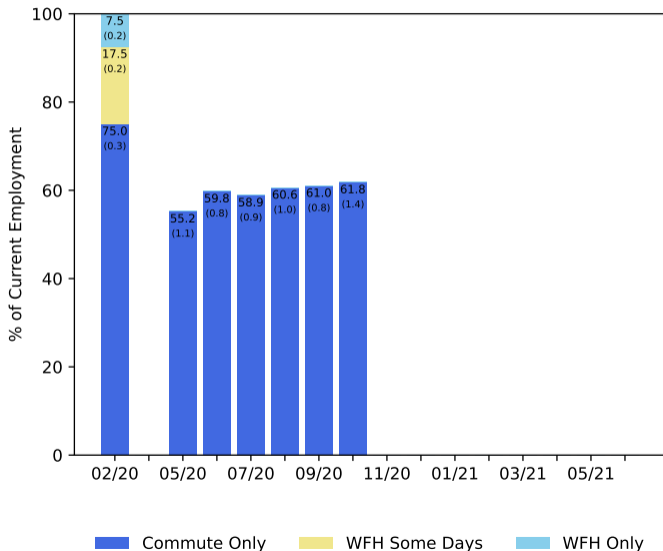
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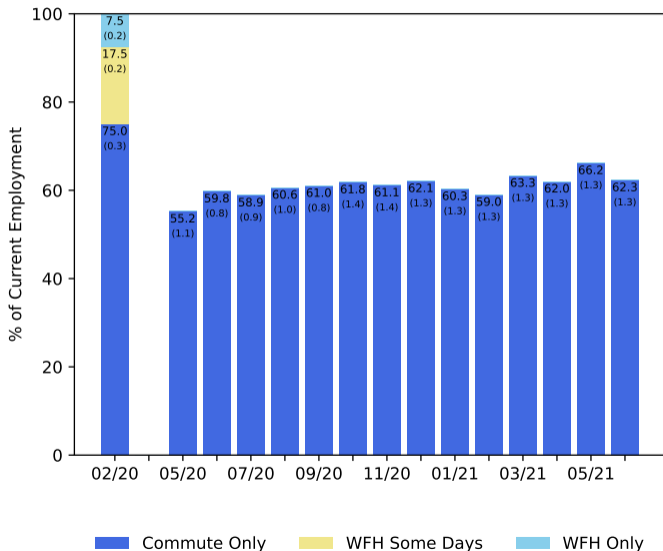
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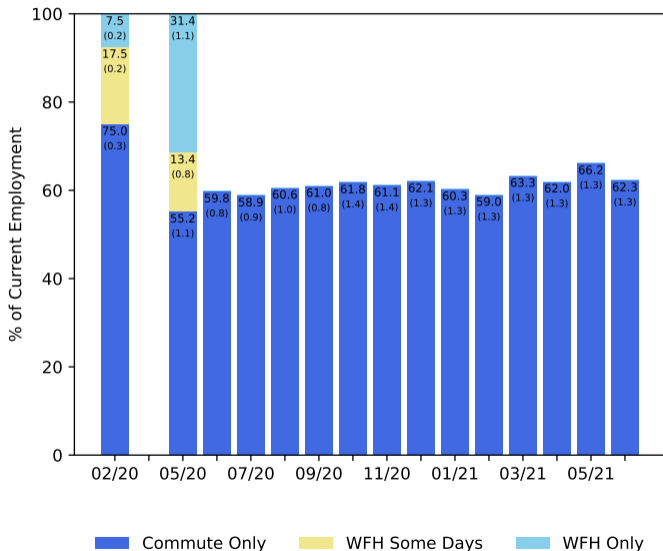
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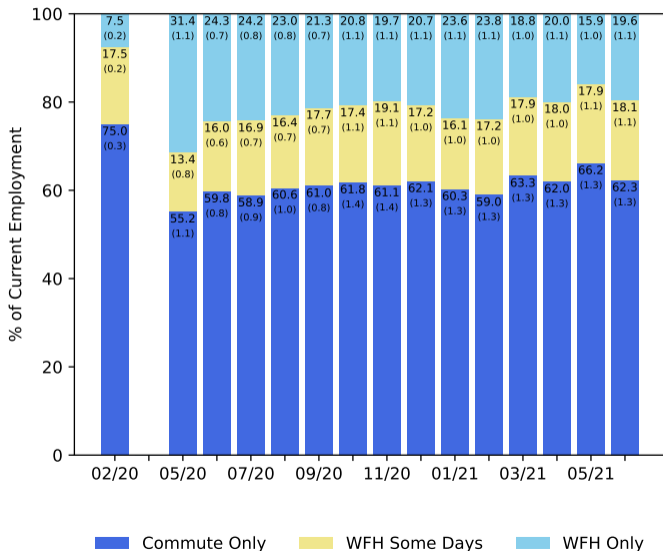
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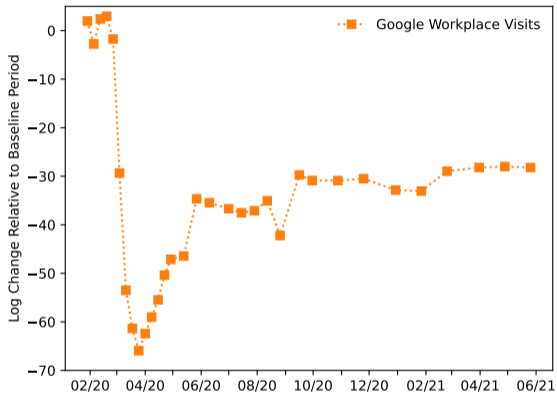
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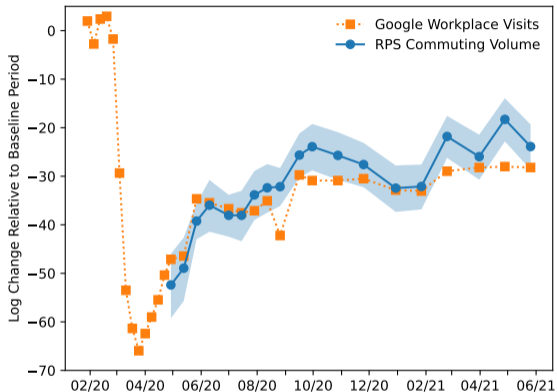
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Comparisons to Other WFH Measures



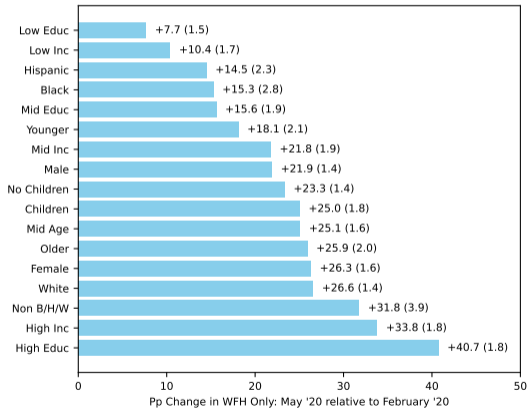
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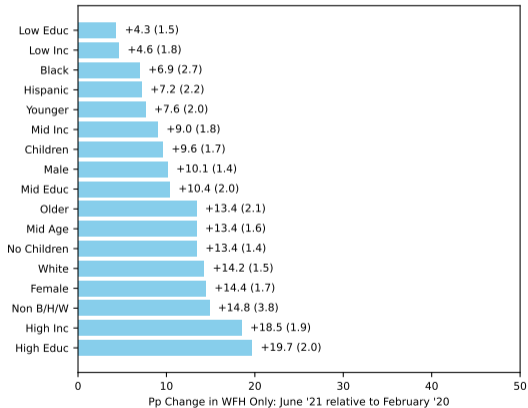
- RPS and Google commuting volume closely align in level and trend
- Feb 20 - June 21 shortfall: more WFH (74.4%), lower employment (21.5%) , shorter workweek (4.1%)

Demographics

May 2020

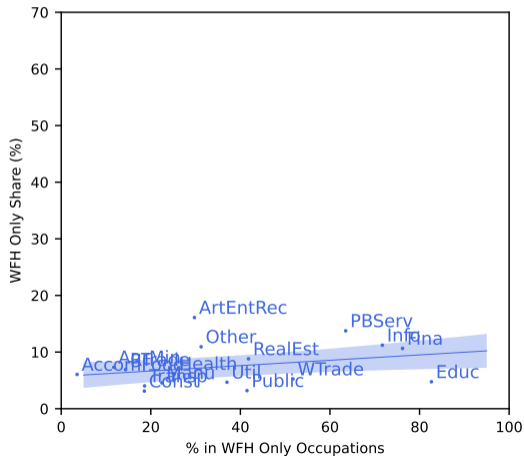


June 2021

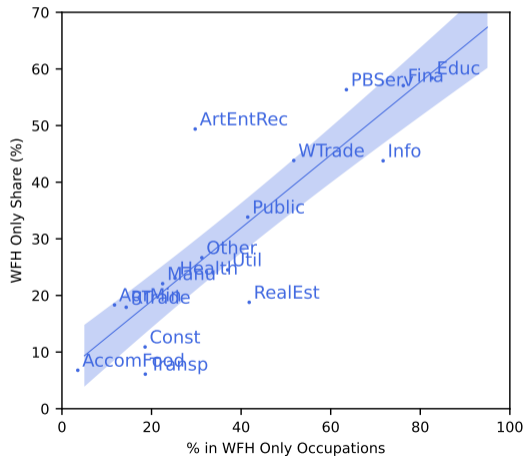


Actual WFH and WFH Ability

Pre-Pandemic (Feb. 2020)

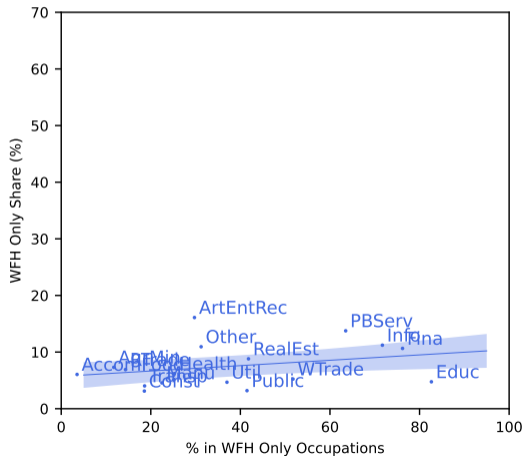


During Pandemic (June 2021)

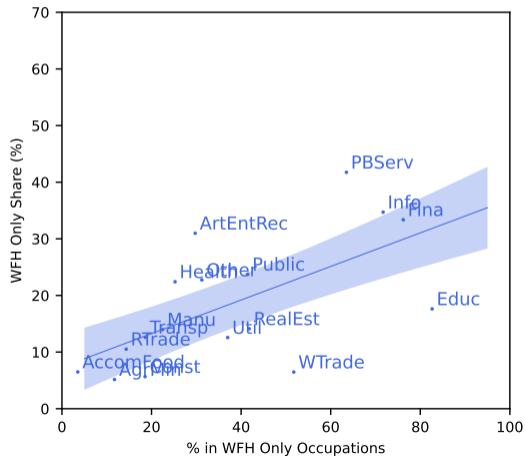


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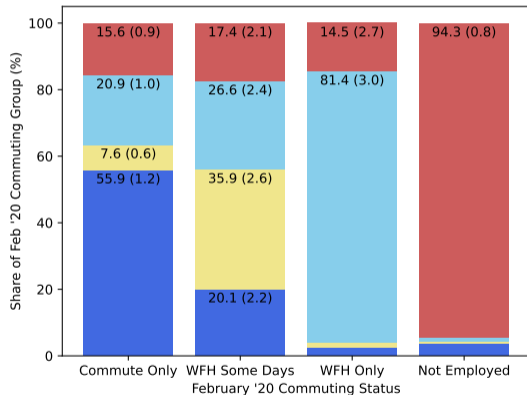


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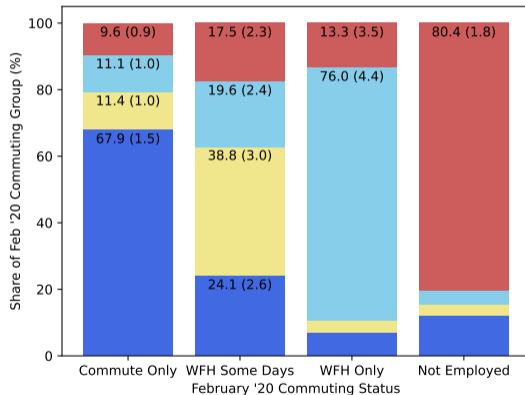


Most WFH Workers Used to Be Full-Time Commuters

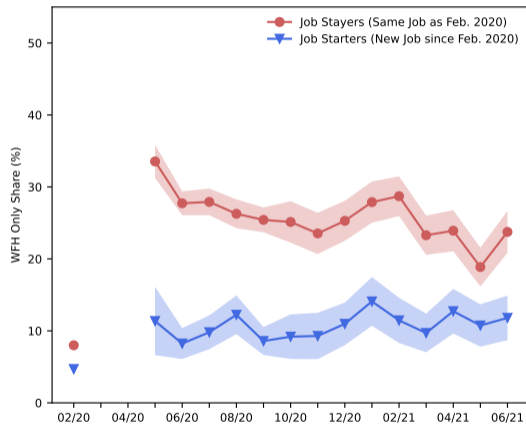
May 2020



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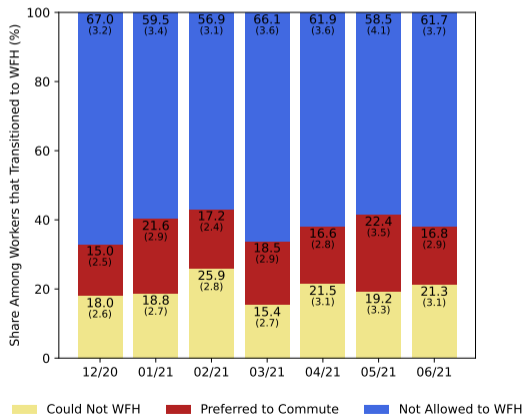
Mostly On-the-Job Transitions



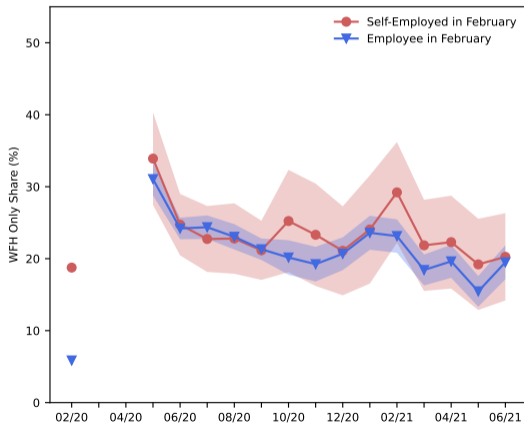
Survey Evidence for WFH Adoption

Job Stayers

Reasons for Commuting in Feb. 2020



Employees vs Self-Employed



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Sketch of Model Environment (1/2): Workers

- Workers have linear preferences $u(w, l, h) = w - h - (1 + \chi)l$
 - ▶ w , wage
 - ▶ work at home h or on-site l
 - ▶ $\chi > 0$, costs associated with on-site work

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- Choices
 - ▶ Individual works if: $w_l \geq 1 + \chi$ or $w_h \geq 1/z$
 - ▶ Workers commute if: $w_l - (1 + \chi) \geq w_h - 1/z$

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 - ▶ Commuter wage: $w_l = 1 + \chi$
- Fraction θ of firms hires commuters and remote workers
 - ▶ WFH labor supply: $E_h = \gamma(w_h)^\lambda$
 - ▶ Remote wage: $w_h = \left(\frac{\lambda}{1+\lambda}\right) (1 + \chi)$

WFH Substitution and Adoption in a Pandemic

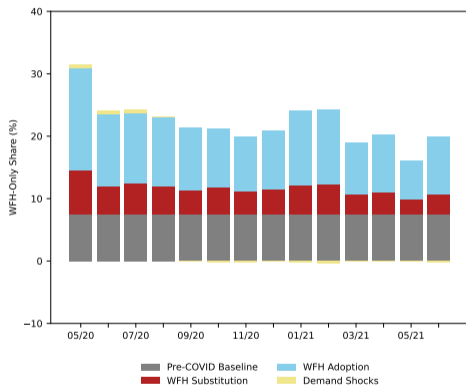
1. WFH **substitution**: pandemic increases on-site costs χ
 - ▶ After pandemic, WFH likely subsides
2. WFH **adoption**: pandemic increases WFH access θ
 - ▶ Pandemic removes constraint on work arrangements
 - ▶ WFH may outlast pandemic since everyone weakly better off

Overview of Quantitative Exercise

- Use model to quantify role of **substitution**, χ , vs. **adoption**, θ
- Treat industries, i as separate labor markets
- Two phases
 - ▶ Pre-pandemic (February 2020), $t = 0$
 - ▶ Pandemic (May 2020 - June 2021), $t = 1, \dots, 15$
- Set $\lambda, \gamma_i, \theta_i^{pre}$ using RPS data, evidence on relative WFH wages and pre-pandemic (WFH) employment levels.
- Set $\chi_{i,t}, \theta_{i,t}, \delta_{i,t}$ to match monthly employment, WFH employment, wages

Model Based Decomposition

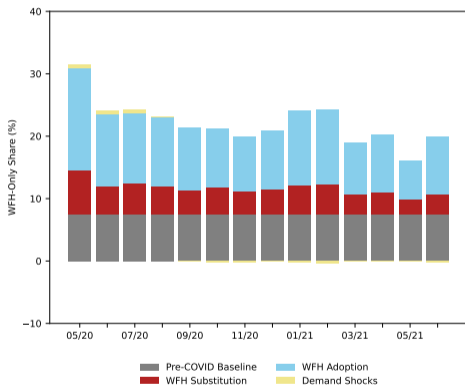
Rise in full-time WFH



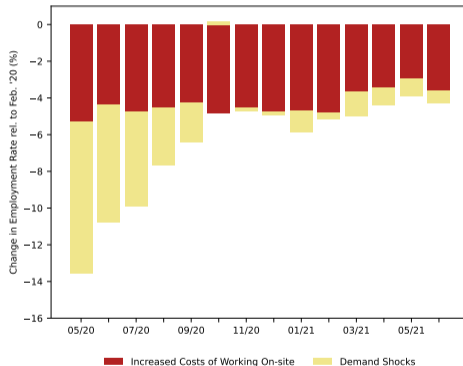
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Model Based Decomposition

Rise in full-time WFH



Employment



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- Suppose demand and costs of working on-site $\delta_{i,t}, \chi_{i,t}$ return to Feb 2020 levels
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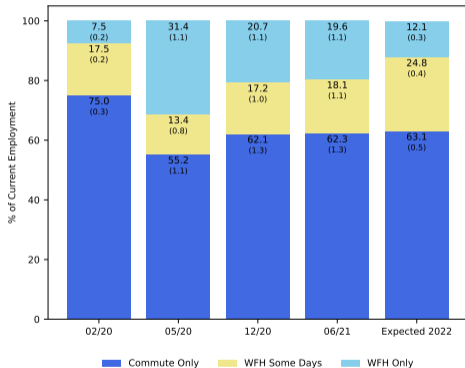
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- Full-time WFH share doubles to 14.6 percent from 7.5 percent
- 51.3 percent has option for at least some WFH, up from 33.4 percent
- Share of WFH workdays rises to 21.3 percent from 14.4 percent

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Expectations for WFH One Year Ahead

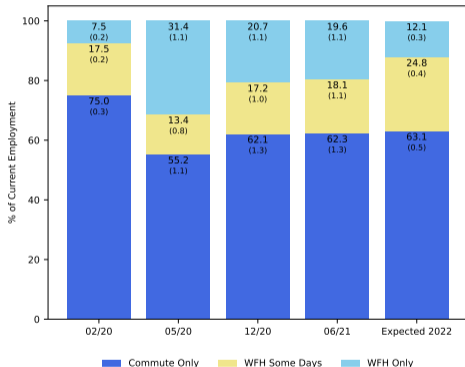
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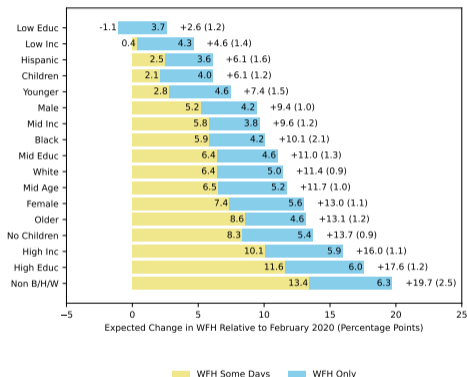
Implied expected share of WFH workdays is 23.4 percent

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Rise in full-time WFH



Employment



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Follow-Up Surveys: February and June 2022

	WFH Status					
	Feb '22			Jun '22		
	Actual	Forecast	Exp.	Actual	Forecast	Exp.
Commute Only	67%	63%	69%	66%	64%	68%
WFH Some Days	20%	25%	21%	19%	24%	20%
WFH Only	14%	12%	10%	14%	12%	11%

- Forecast are the expectations from the Feb '21 and Jun '21 survey, respectively
- Expectations are the one year ahead expectations for Feb '23 and Jun '23, respectively

Conclusion

- Beyond initial months of the pandemic, rise in WFH is mostly due to adoption of WFH arrangements
- Likely upward shift in pre-existing secular trend towards more WFH
- Benefits highly unequally distributed