

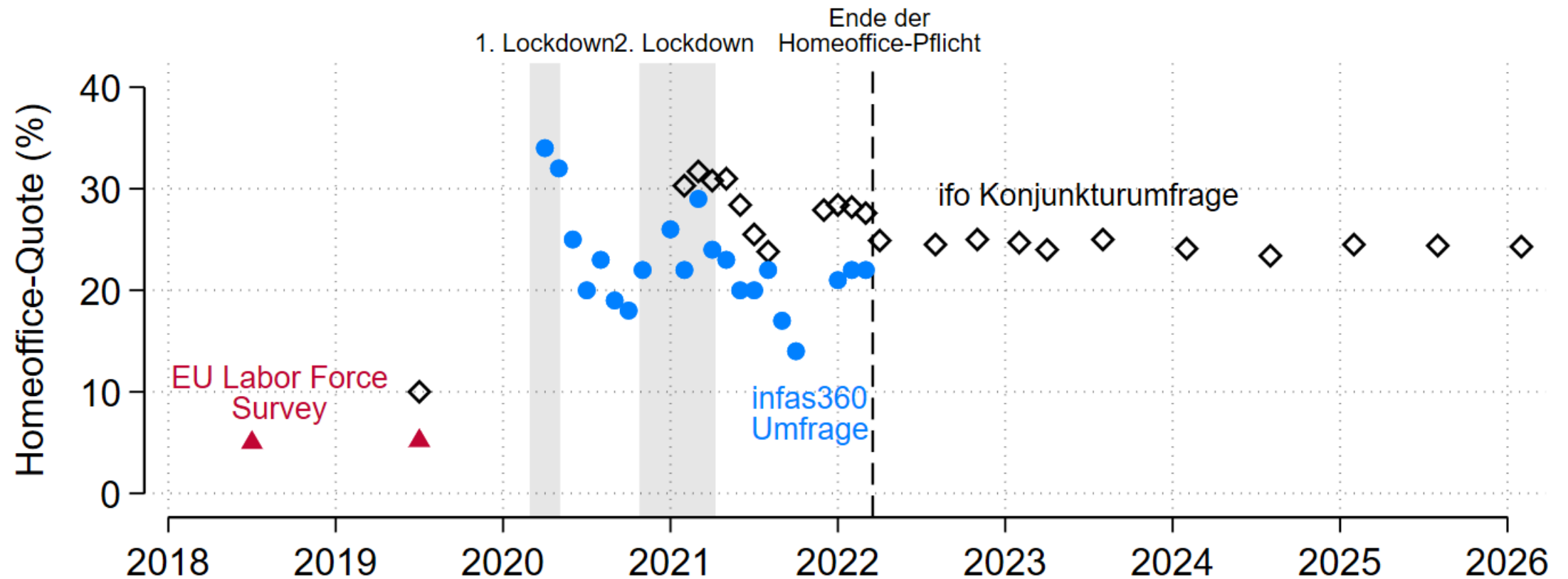
WORKING FROM HOME, URBAN WAGE GROWTH, AND INEQUALITY

Annekatriin Niebuhr (Kiel University, IAB), Jan Cornelius Peters (Thünen
Institute), Anja Rossen (IAB), Duncan Roth (IAB, IZA, CESifo)

Cambridge Journal of Regions, Economy and Society 2026 conference,
St. Catharine's College, Cambridge, UK
16 - 17 of July, 2026

Motivation

Increase in working from home



Anmerkungen: Anteil Beschäftigte, die mind. teilweise Homeoffice nutzen.
© J.V. Alipour, ifo Institut.

Motivation

- Persistent increase in working from home (WFH) caused by COVID-19 pandemic $\Rightarrow$ spatial implications: changes in urban life and economic activity in cities (Glaeser 2022, Nathan & Overman 2020, Florida et al. 2023)
- Evidence on commuting/migration (Coskun et al. 2026, Knuepling et al. 2025, Li & Su, 2026), housing/commercial real estate (Behrens et al. 2024, Liu & Su 2021), spatial distribution of spending (Alipour et al. 2022, Althoff et al. 2022)
- Little evidence on how WFH affects size of agglomeration economies – might weaken agglomerations economies related to face-to-face interaction (Liu & Su 2025, Gaetani et al. 2026)

Literature

- Urban wage premium: workers in large cities earn significantly more than workers in rural areas (Glaeser & Maré 2001, Wheeler 2006, Yankow 2006) – due to
 - Sorting of more productive workers into large cities (Combes et al. 2008)
 - Agglomeration effects, workers at top of wage distribution seem to benefit above average (Carlsen et al. 2016, De La Roca & Puga 2017) $\Rightarrow$ wage inequality
- Urban wage growth premium: learning advantages in large cities as important source of productivity/wage growth (Marshall 1890, Glaeser 1999, Duranton & Puga 2004)
 - knowledge exchange via opportunities for face-to-face meetings and high rate of (new) contacts between workers, both increase with city size (Glaeser 1999, Storper & Venables 2004, Berliant et al. 2006)
 - higher wage premium for urban than for rural work experience, even after moving to less dense locations (De La Roca & Puga 2017, Peters 2020)

Literature

- Increase in WFH after pandemic might reduce knowledge exchange due to less face-to-face contact of workers (Yang et al. 2022, Emanuel et al. 2026)
- Effects of WFH on size of agglomeration economies
 - Liu and Su (2025): 18.4% decrease in urban wage premium in U.S. for occupations with high levels of WFH during pandemic, reduced wage returns to interpersonal skills
 - Gaetani et al. (2026): innovation premium of U.S. cities declined during pandemic by 18.5% up to 22.9% relative to pre-pandemic level, smartphone data suggest that important part of drop is due to a decline in frequency of local interactions
- Role of agglomerations economies for increase of inequality between and within regions
 - Increase in returns to size due to structural change from manufacturing to digital technologies (Bathelt et al. 2024, Kemeny & Storper 2024)
 - Increasing differences in urban wage premium between low-/medium- and high-skilled workers (Autor 2019, Baum-Snow et al. 2018), role of learning (Baum-Snow et al. 2018)

Main research questions

Increase in WFH likely results in fewer (accidental) meetings and less interaction between workers at the workplace

⇒ Weaken learning benefits that large cities offer

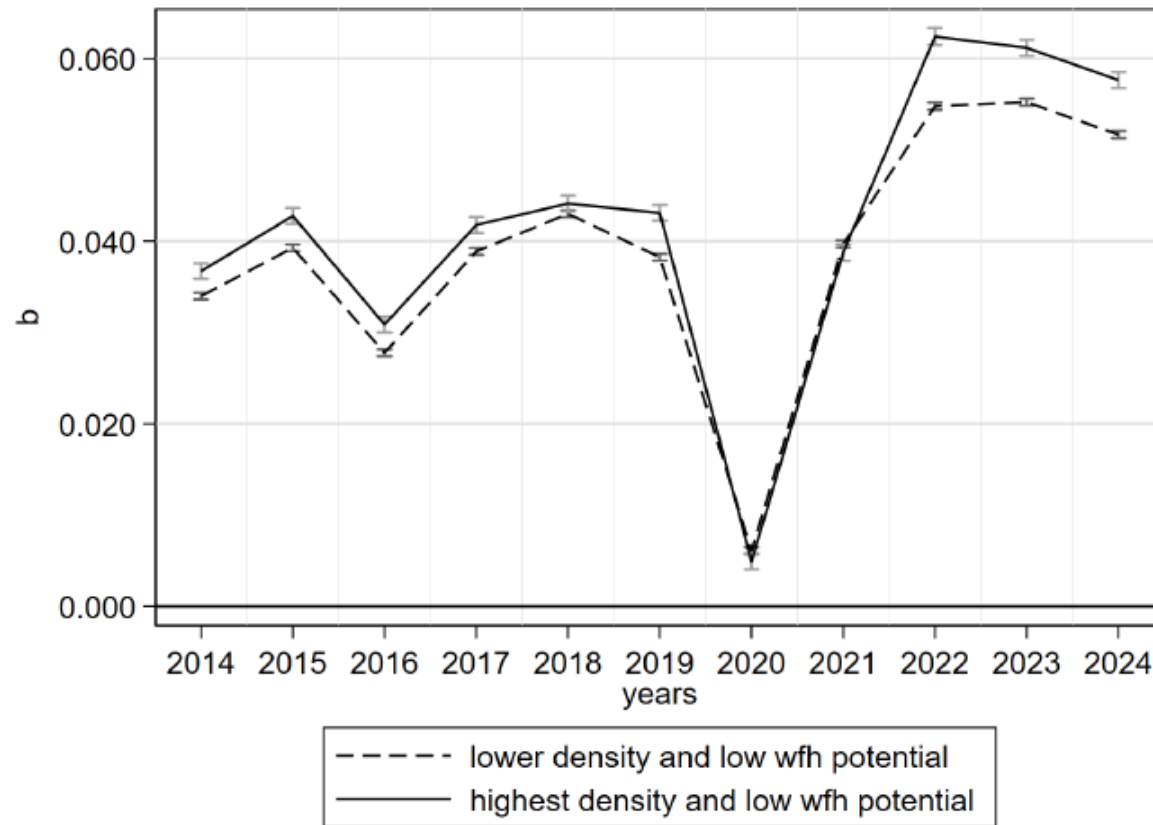
⇒ Decline in wage growth in cities relative to less densely populated areas

- (1) Does the urban wage growth premium decline after the pandemic-related increase in remote work in 2020?
- (2) Do mainly workers in occupations with a high capacity to work from home suffer from a decline in the urban wage growth premium?
- (3) What are the implication for inequality between and within regions?

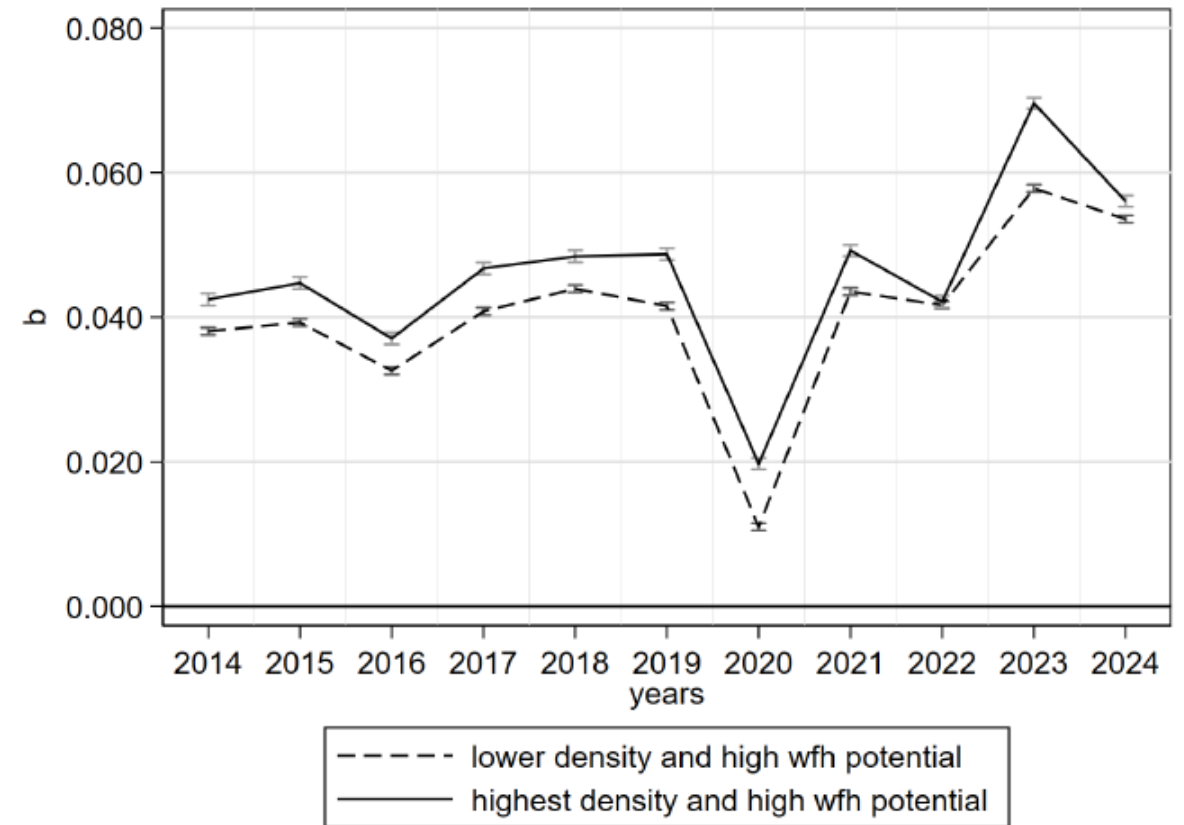
Data

- Integrated Employment Biographies (IAB) - 10 % random sample of all workers employed in Germany 2013–2024
 - worker-year panel referring to June 30, full-time employment outside public sector, temporary agency employment
 - 17.5 mill. worker-year observations for 2.5 mill. individuals
- Information: gross daily wage, occupation, establishment (municipality, size, sector), work experience since labour market entry, establishment tenure
- Pivotal variables
 - wage growth between year t and $t - 1$
 - location of work in $t - 1$: highest density vs. lower density locations [[map](#)]
 - work experience gained between June 30, t and $t - 1$, by location type
- Capacity to work from home in 2018 by occupation (Alipour et al. 2023) [[figure](#)]
- Monthly information on short-time work 2009–2024 on establishment level (Kagerl 2026)

Average annual wage growth (unconditional)



(a) Low working-from-home capacity



(b) High working-from-home capacity

Estimation strategy

$$\begin{aligned}\Delta \ln(wage)_{i,r,t} = & \beta I(r \in \text{HD}) + \sum_{\substack{\tau \in [2014, 2024], \\ \tau \neq 2019}} \delta_{\tau} I(t = \tau) I(r \in \text{HD}) \\ & + u_i + \gamma_t + \phi_{o(i,t),t} + \mu_{s(i,t),t} + \theta_{f(i,t),t} + X_{i,t} \rho_t + \varepsilon_{i,t}\end{aligned}$$

β, δ_{τ} : urban wage growth premium and its variation over time (relative to 2018→2019)

$r(i, t)$: municipality of work of worker i in year t (establishment location)

u_i, γ_t : FE for worker i and year t , respectively

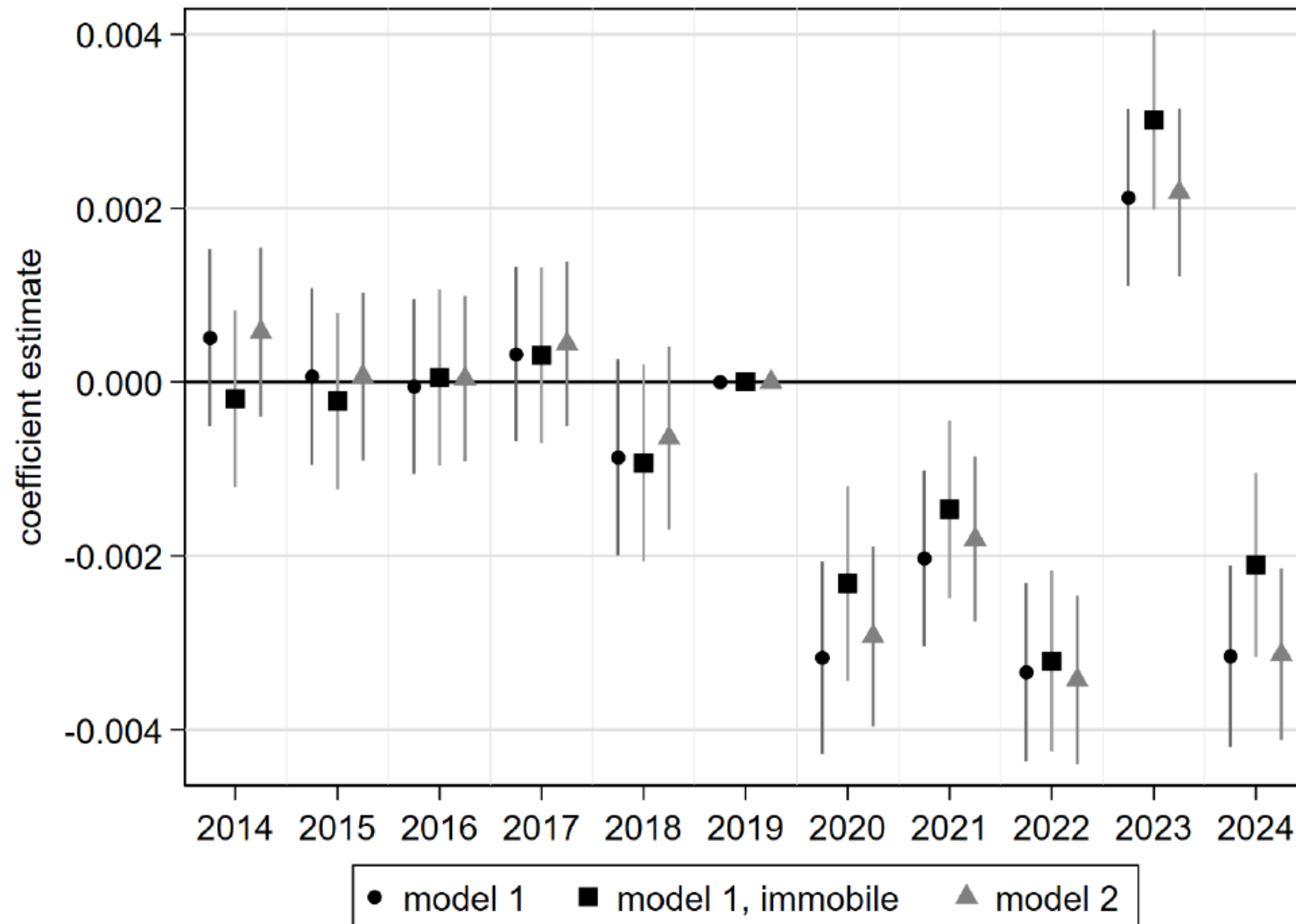
$\phi_{o(i,t),t}$: FE for occupation $\times$ year

$\mu_{s(i,t),t}$: FE for sector $\times$ year

$\theta_{f(i,t),t}$: FE for establishment size category $\times$ year

$X_{i,t}$: total work experience and establishment tenure of worker i in year t , indicators for moving to / away from highest density / establishment change between t and $t + 1$

Changes in relative wage growth (conditional) – highest versus lower density locations



Differences between workers with high and low WFH capacity

Differences between age groups [\[table\]](#)

	all workers		working from home capacity			
			low		high	
	(1)	(2)	(3)	(4)	(5)	(6)
Highest density	0.002*** (0.00)	0.002*** (0.00)	0.003*** (0.00)	0.003*** (0.00)	0.003*** (0.00)	0.003*** (0.00)
Highest density x year						
2014	0.001 (0.00)	0.001 (0.00)	0.001 (0.00)	0.001 (0.00)	0.000 (0.00)	0.000 (0.00)
2015	0.000 (0.00)	0.000 (0.00)	-0.000 (0.00)	0.000 (0.00)	0.001 (0.00)	0.001 (0.00)
2016	-0.000 (0.00)	0.000 (0.00)	0.000 (0.00)	0.000 (0.00)	-0.000 (0.00)	-0.000 (0.00)
2017	0.000 (0.00)	0.000 (0.00)	-0.000 (0.00)	-0.000 (0.00)	0.001* (0.00)	0.001* (0.00)
2018	-0.001 (0.00)	-0.001 (0.00)	-0.001 (0.00)	-0.001 (0.00)	-0.001 (0.00)	-0.000 (0.00)
2019	– reference –					
2020	-0.003*** (0.00)	-0.002*** (0.00)	-0.006*** (0.00)	-0.002*** (0.00)	-0.001 (0.00)	-0.002* (0.00)
2021	-0.002*** (0.00)	-0.001* (0.00)	-0.003*** (0.00)	-0.001 (0.00)	-0.001 (0.00)	-0.001 (0.00)
2022	-0.003*** (0.00)	-0.003*** (0.00)	0.001 (0.00)	-0.001 (0.00)	-0.007*** (0.00)	-0.005*** (0.00)
2023	0.002*** (0.00)	0.001*** (0.00)	0.001 (0.00)	-0.000 (0.00)	0.004*** (0.00)	0.003*** (0.00)
2024	-0.003*** (0.00)	-0.004*** (0.00)	-0.000 (0.00)	-0.000 (0.00)	-0.005*** (0.00)	-0.005*** (0.00)
N	17,526,623	17,526,623	9,507,883	9,507,883	7,922,653	7,922,653
R²	0.119	0.120	0.133	0.134	0.116	0.116
Workers	2,461,367	2,461,367	1,425,699	1,425,699	1,150,321	1,150,321
Conditional on short-time work	no	yes	no	yes	no	yes

Implications for inequality: differences in WFH capacity across skill groups

	wfh capacity			density category		
	low	high	total	lower	highest	total
low-skilled	73 %	27 %	100 %	72 %	28 %	100 %
medium-skilled	62 %	38 %	100 %	78 %	22 %	100 %
high-skilled	20 %	80 %	100 %	58 %	42 %	100 %
unknown	79 %	21 %	100 %	71 %	29 %	100 %
all	52 %	48 %	100 %	72 %	28 %	100 %

- In addition: relatively strong decline of urban wage growth premium for high-skilled with high WFH capacity (compared to medium-skilled workers with high WFH capacity, less clear for low-skilled) [[table](#)]
- Workers with high WFH capacity $\Rightarrow$ high worker FE, occupation FE, municipality FE, location type FE [[figure](#)]
- High-skilled, high-income workers more exposed to increase in WFH / decline of face-to-face interaction, stronger decline in urban wage growth premium

Conclusions

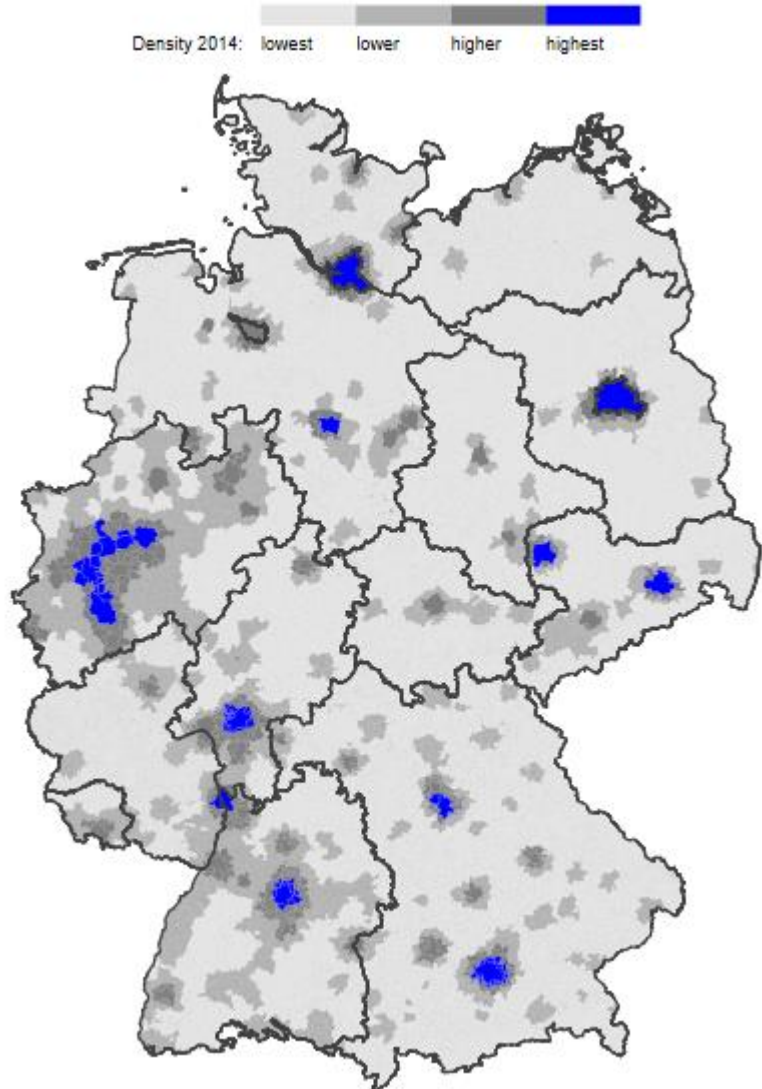
- Significant decrease in urban wage growth premium after increase in WFH: on average by about 0.2 percentage points lower for 2020 to 2024 compared to period before 2020
- Mainly for workers with high WFH capacity, clear pattern for age group 25-44, that works most frequently remote
- Might point to decline in agglomeration economies due less face-to-face interaction between workers $\Rightarrow$ decline in learning benefits of large cities
- High-skilled workers tend to show high WFH capacity and concentrate in large cities
- Wage growth reduced primarily for high-income workers employed in high-income regions $\Rightarrow$ increase in WFH might result in declining disparities between regions and within large cities

THANK YOU FOR YOUR ATTENTION

Annekatrien Niebuhr

annekatrien.niebuhr@iab.de

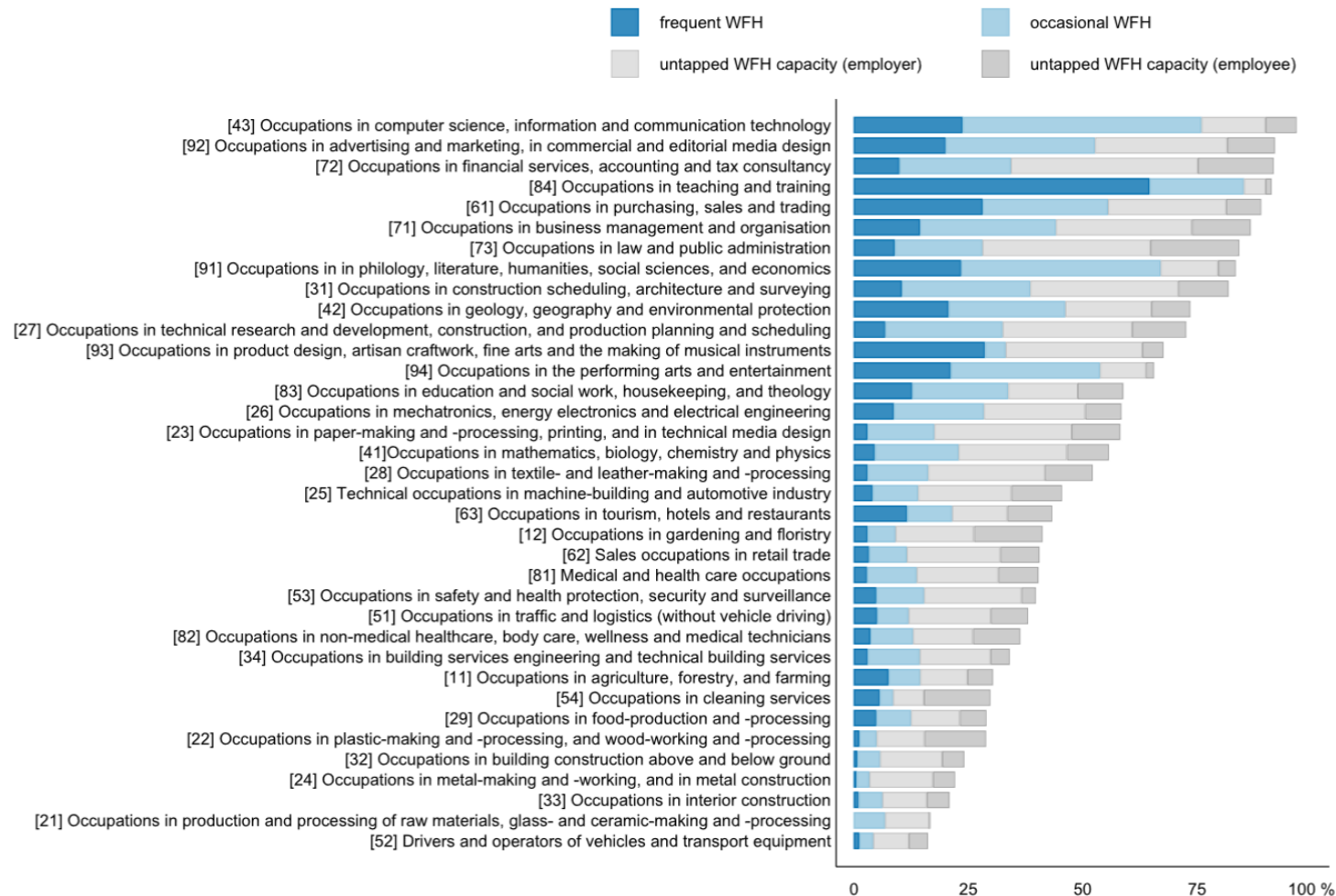
Categories of local employment density



Each category accounts for about 25% of total employment in Germany, local employment density refers to employment within a 10 km radius of the municipal centre

[\[BACK\]](#)

Capacity to work remotely by occupation, 2018 (Alipour et al. 2023)



[\[BACK\]](#)

Fig. 1. Capacity to Work from Home by Occupation

Notes: The figure displays WFH capacities and pre-pandemic WFH usage by occupation (2-digit KldB). The underlying values are reported in [Table A.1](#).

Source: 2018 BIBB/BAuA Employment Survey, own calculations.

Differences across age groups, workers with high WFH capacity

[\[BACK\]](#)

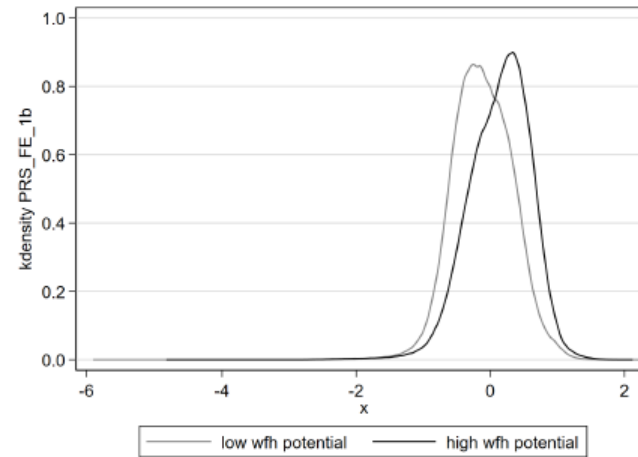
	<25 years	<25 years const. edu.	25–44 years	>44 years
	(1)	(2)	(3)	(4)
Highest density	0.006 (0.01)	0.003 (0.00)	0.004*** (0.00)	-0.000 (0.00)
Highest density x year				
2014	0.022*** (0.01)	0.017*** (0.01)	0.001 (0.00)	-0.002** (0.00)
2015	0.020*** (0.01)	0.016*** (0.01)	-0.000 (0.00)	0.001 (0.00)
2016	0.014*** (0.00)	0.013*** (0.00)	-0.001 (0.00)	-0.000 (0.00)
2017	0.009** (0.00)	0.008** (0.00)	0.001 (0.00)	0.001 (0.00)
2018	0.003 (0.00)	0.005 (0.00)	-0.001 (0.00)	-0.000 (0.00)
2019		– reference –		
2020	-0.003 (0.00)	-0.000 (0.00)	-0.002 (0.00)	-0.000 (0.00)
2021	-0.003 (0.00)	0.000 (0.00)	-0.001 (0.00)	0.001 (0.00)
2022	0.003 (0.00)	0.008* (0.00)	-0.003** (0.00)	-0.006*** (0.00)
2023	-0.006 (0.01)	-0.000 (0.00)	-0.002 (0.00)	0.010*** (0.00)
2024	-0.026*** (0.01)	-0.022*** (0.01)	-0.007*** (0.00)	0.000 (0.00)
N	343,208	333,906	3,921,496	3,544,587
R²	0.344	0.341	0.142	0.084
Workers	112,031	109,751	690,490	530,202

Differences across skill groups

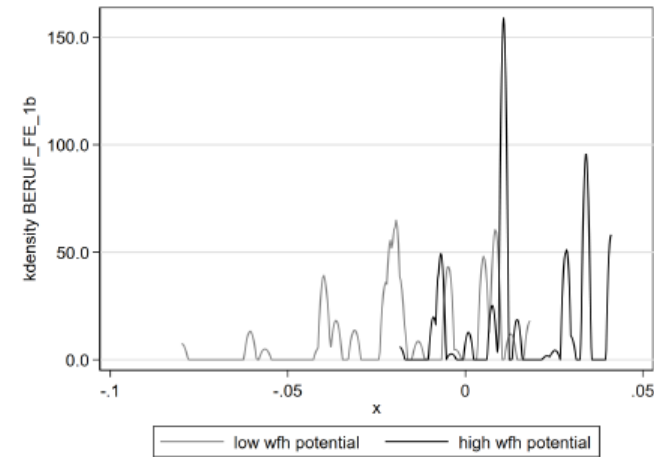
[\[BACK\]](#)

	low-skilled		medium-skilled		high-skilled	
	wfh capacity		wfh capacity		wfh capacity	
	low (1)	high (2)	low (3)	high (4)	low (5)	high (6)
Highest density	0.003 (0.004)	0.009 (0.007)	0.006*** (0.001)	0.001 (0.001)	0.002 (0.003)	0.005*** (0.002)
Highest density x year						
2014	-0.001 (0.004)	-0.006 (0.007)	0.002 (0.001)	0.002 (0.001)	-0.004 (0.004)	0.000 (0.002)
2015	0.003 (0.004)	-0.010 (0.007)	-0.001 (0.001)	-0.000 (0.001)	-0.002 (0.004)	0.000 (0.002)
2016	-0.000 (0.004)	-0.002 (0.007)	0.001 (0.001)	0.000 (0.002)	-0.005 (0.004)	-0.001 (0.002)
2017	-0.002 (0.004)	0.003 (0.007)	-0.000 (0.001)	0.001 (0.001)	0.001 (0.004)	0.001 (0.002)
2018	0.001 (0.004)	-0.013** (0.007)	-0.002 (0.001)	0.001 (0.002)	-0.002 (0.004)	-0.001 (0.002)
2019			– reference –			
2020	-0.007 (0.005)	-0.008 (0.007)	-0.003** (0.001)	0.000 (0.002)	-0.006 (0.005)	-0.002 (0.002)
2021	0.003 (0.005)	-0.002 (0.006)	-0.001 (0.001)	0.002 (0.002)	-0.001 (0.004)	-0.004** (0.002)
2022	0.003 (0.004)	-0.008 (0.006)	-0.001 (0.001)	0.002 (0.002)	-0.003 (0.004)	-0.006*** (0.002)
2023	-0.000 (0.004)	-0.002 (0.006)	-0.000 (0.001)	0.001 (0.002)	-0.000 (0.004)	-0.003 (0.002)
2024	-0.005 (0.004)	-0.008 (0.006)	-0.002 (0.001)	-0.003** (0.002)	0.000 (0.004)	-0.009*** (0.002)
N	328,613	113,561	3,449,653	1,945,184	472,713	1,667,811
R²	0.186	0.175	0.149	0.157	0.147	0.131
Workers	65,051	22,155	590,051	339,383	96,284	293,090

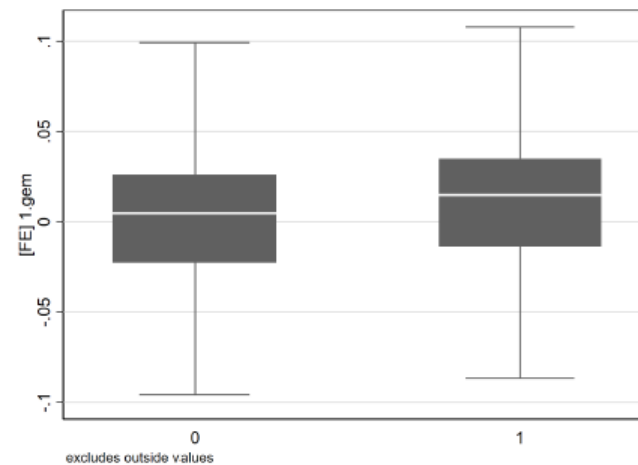
Differences in WFH capacity across fixed effects: worker, occupations, municipality/location type



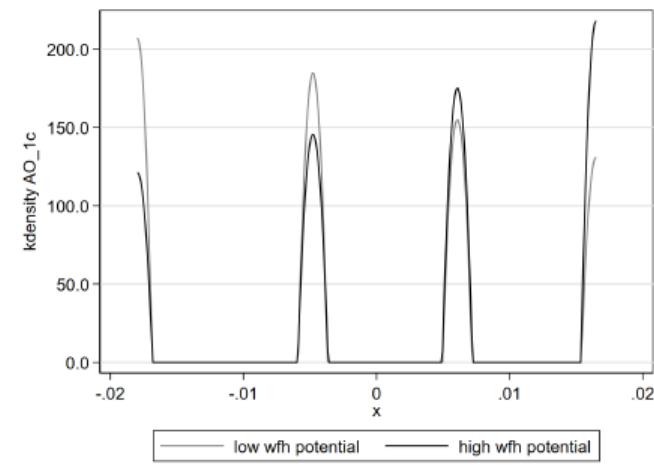
(a) x-axis: worker FE



(b) x-axis: occupation FE



(c) x-axis: WFH capacity, y-axis: municipality FE



(d) x-axis: location type FE

[\[BACK\]](#)