

Dormant skills as a result of regional and skill mismatch?

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Motivation

- Skill shortages
- High unemployment

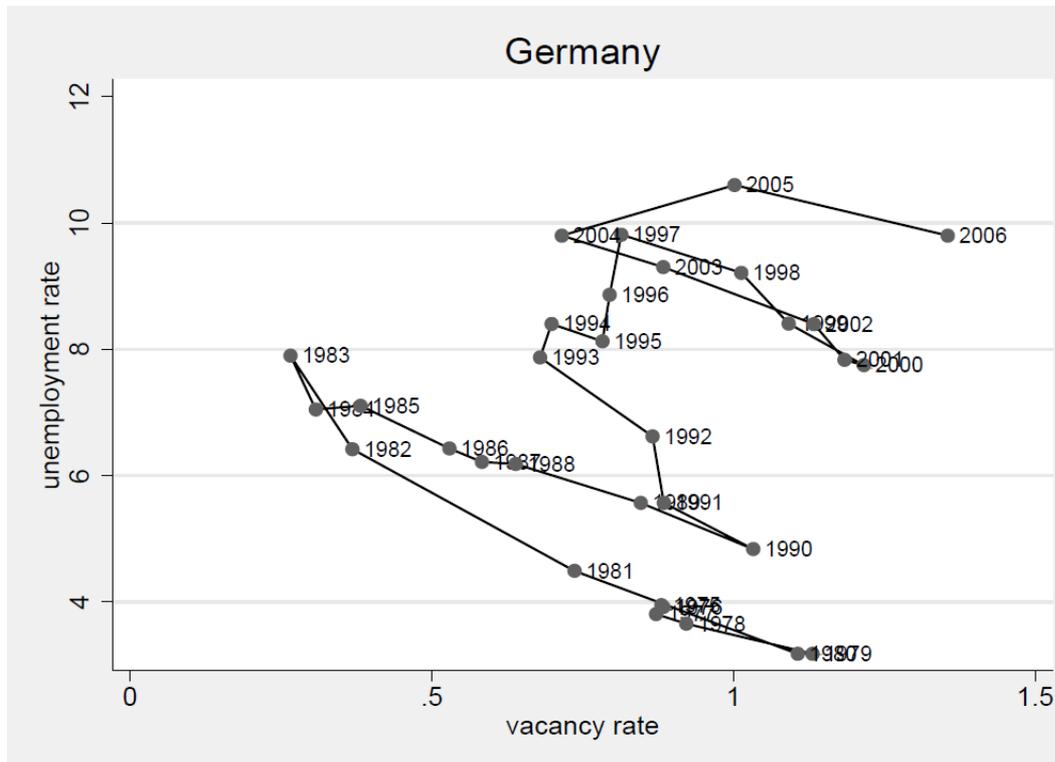


Figure 1. The Beveridge curve 1976-2006 (Source: Bouvet 2009)

Why do shortages and unemployment coexist?

- Asymmetric information
- Coordination failures
- Institutional arrangements
- Restricted regional labour mobility (regional mismatch)
- Heterogeneity in the demand and supply of skills (skill mismatch)
(Pissarides 2000, Petrongolo and Pissarides 2001)
- Political relevance
 - Skills mismatch → requalification policy
 - Regional mismatch → labour mobility or other policy

Regional mismatch

- Are the jobs where the unemployed skill-stocks are?
- Low/medium skilled unemployed/jobs: less regionally mobile than high-skilled (OECD 2008, Mauro and Spilimbergo 1998)
- Do unemployed migrate to regions where the jobs are?

H1: The geographical distance to the job openings has a negative effect on the exit from unemployment.

Skill mismatch

- Demand side: changing skill demands due to:
 - Demographic changes (health-care jobs),
 - International division of labour (textile jobs),
 - Technological change (complementing creative and substituting routine skills).
- Supply side: inelastic supply for certain occupations due to the duration of training and retraining.
- Human capital is transferable across occupations (Poletaev and Robinson 2008, Gathmann and Schönberg 2010).
- Among the unemployed, those trained in *occupations skill-related* to those in demand should have an advantage in the reemployment.

Skill mismatch

H3: The occupation-specific unemployment rate does not only depend on the demand for the own occupation, but also on the demand for *skill-related* occupations.

H4: The effect is more pronounced for the low-skilled.

H5: The effect is more pronounced for the young workers.

Data

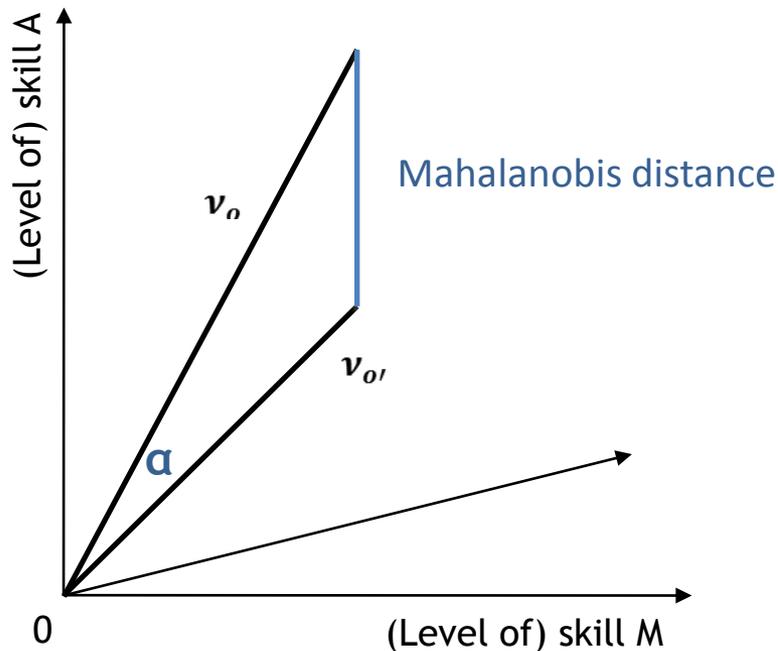
- Occupational and Career Survey (BauA and BIBB, 2005/2006)
 - 20,000 individuals
 - 52 questions on tasks and working conditions
 - Occupations, 3-digit KldB88
- IAB Employment samples (IAB, 1975-2004)
 - 2% sample of the German population subject to social security
 - Employment and unemployment history, occupational history, labour mobility, regional information
- Job openings and vacancy time (Federal Employment Agency, 2000-2009)
 - By 3-digit KldB88 and 16 German states
- Final sample: 120 occupations, 16 states, 2000-2004.

Measurement: skill demand, skill shortages, skill stocks

- Demand: job announcements by occupation and state.
- Shortage: vacancy time.
 - $\text{Corr}(\text{vacancy time, unemployment})=0.71$;
 - $\text{Corr}(\text{vacancy time, real wages})=0.33$;
 - $\text{Corr}(\text{vacancy time, real wage growth})=0.15$.
- Skill stocks: skills required at the last occupation before unemployment.
- Regional distance (same state, bordering state, non-bordering state)

Measurement: skill-relatedness skill-distance of occupations

- Polataev and Robinson (2008), Gathmann and Schönberg (2010), Neffke (2009), Nedelkoska and Neffke (2010), Rocher (2010).



Orthogonal skills (Angular distance, Euclidean distance):
Factor analysis of the 52 tasks
6 skill dimensions:

- cognitive
- manual
- technical
- care
- service
- security

Skill-relatedness

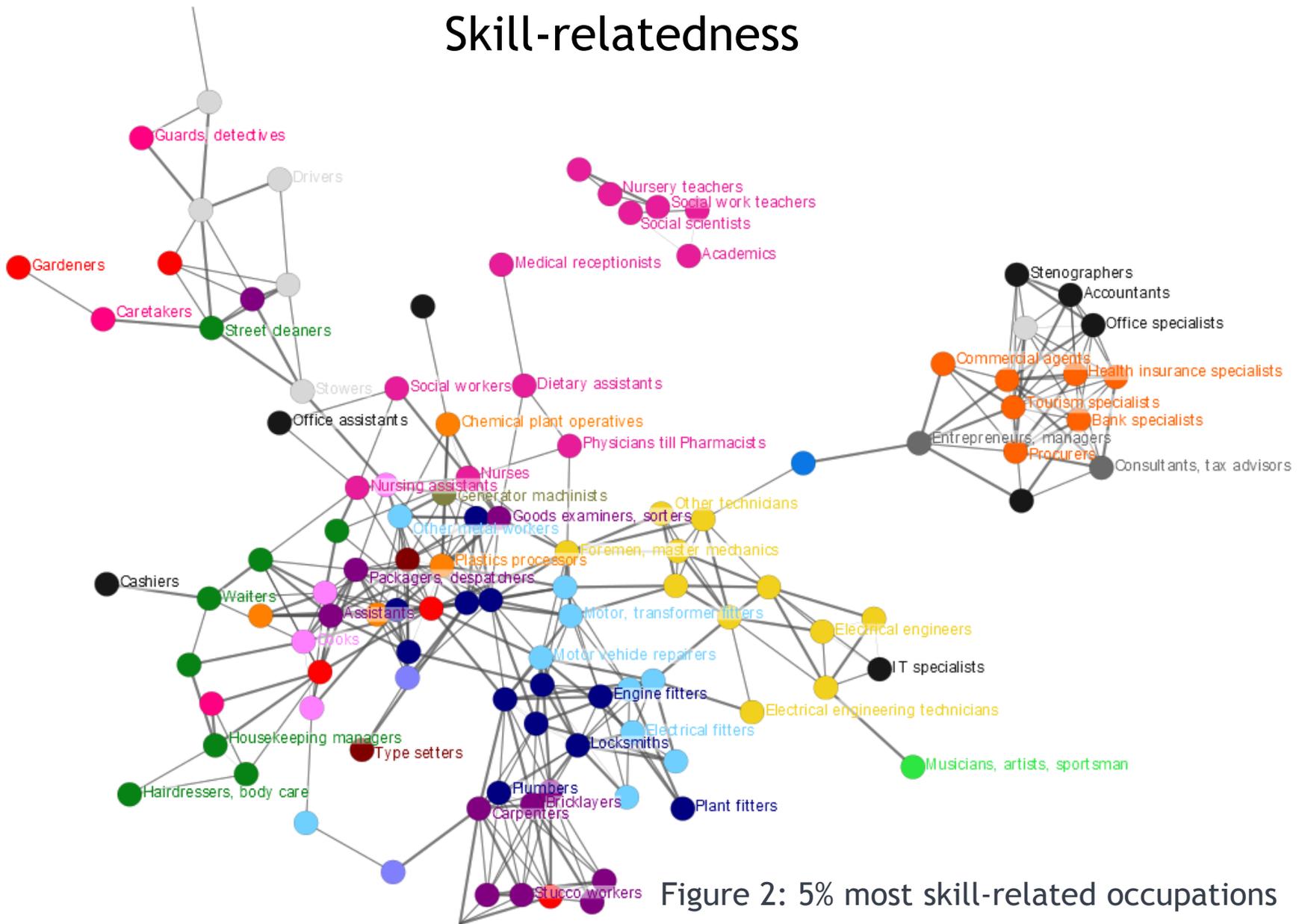


Figure 2: 5% most skill-related occupations

Job openings which are difficult to fill

Occupation	Average vacancy time (2000-2004)
Electrical engineers	96.74
Health insurance specialists	95.38
Bank specialists	94.45
Business dealers	92.78
Mechanical, motor engineers	86.70
Other wood and sports equipment makers	84.54
Toolmakers	84.20
Physicians till Pharmacists	81.74
Electrical engineering technicians	81.68
Postal deliverers	81.44
Conductors	79.94
Other engineers	78.44
Architects, civil engineers	76.98
Other mechanics	75.76
Drillers	74.79
Motor, transformer fitters	72.72
Telecommunications mechanics	71.86
Machinery repairers	71.59
Non-medical practitioners	69.48
Mechanical engineering technicians	69.42
Precision fitters	69.20
Nurses	69.00
Chemists, physicists, mathematicians	66.33
Other technicians	66.04
Nursing assistants	65.82

Source: Federal
Employment Agency, 2011

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Employment Agency, 2011

Skills of the difficult-to-fill jobs

Mechanics	
Handicraft knowledge	0.77
Technical knowledge	0.76
Measuring, checking, proving	0.75
Repair and maintenance of machines	0.71
Work with computers	0.54
Production of goods	0.42
Consult and inform	0.41
Controlling and operating machines/technological processes	0.38
PC software knowledge	0.33

Healthcare occupations	
Consult, inform	0.86
Information collection, research, inform	0.75
Work with computers	0.74
Medical knowledge	0.73
Care, assist, cure	0.69
Advanced knowledge of German	0.46
Measuring, checking, proving	0.45
Organizing, planing, and preparing	0.42
Knowledge in natural sciences	0.37

Regional mismatch: are there dormant skills?

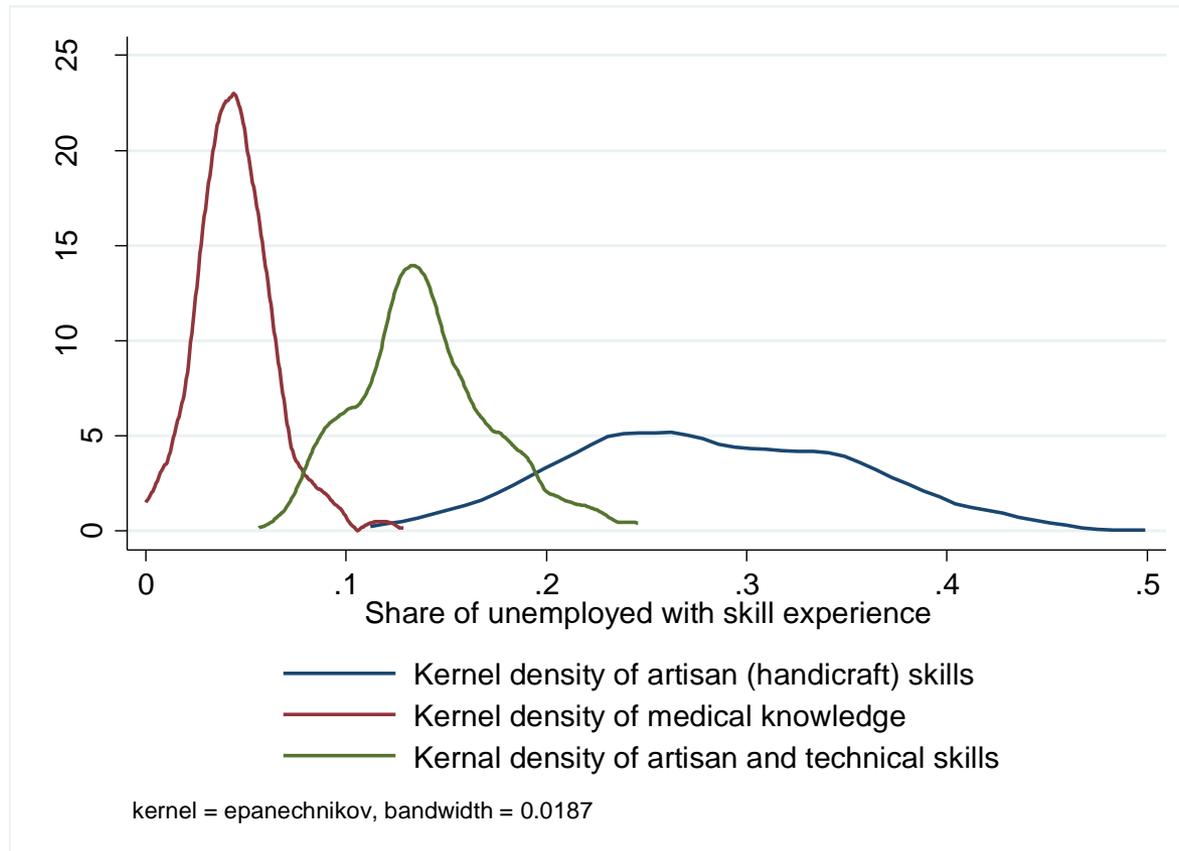


Figure 3: Share of unemployed with experience in different skills

Is there a regional mismatch?

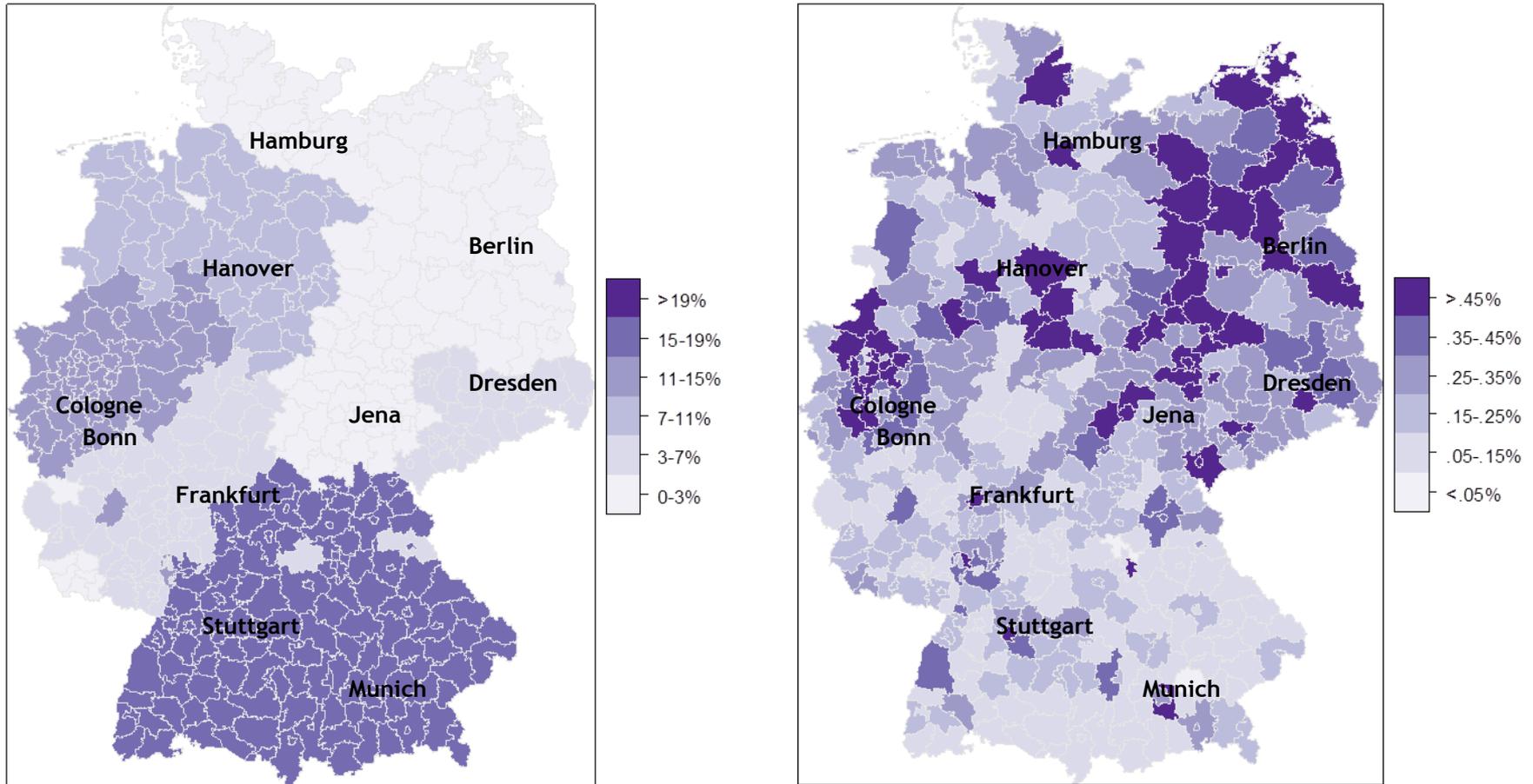


Figure 4: Regional distribution of artisan skill experience (left)
Regional distribution of artisan+technical skill experience (right)

Regional mismatch: does labour migrate to job creating regions?

$$move_{olo'l'} = \beta_0 + \beta_1 RD_{ll'} + \beta_2 openings_{o'l'} + \beta_3 RD * openings_{o'l'} + \beta_4 De_{oo'} + \beta_5 unempl_{ol} + \beta_6 w_{o'l'} + \beta_7 size_{o'l'} + \varepsilon_{olo'l'}$$

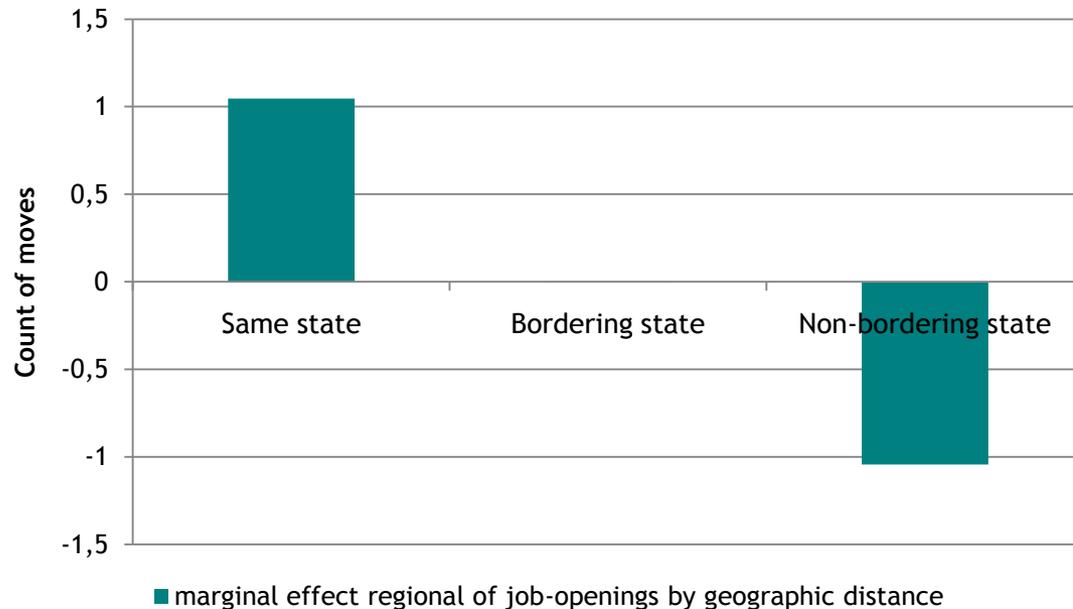


Figure 5: Interacion effects openings share*regional distance

Skill mismatch: employing from related occupations?

$$\Delta unempl_{ol} = \beta_0 + \beta_1 openings_{ol} + \beta_2 openings_{rl} + \beta_3 \Delta w_{ol} + \varepsilon_{ol}$$

H1: The occupation-specific unemployment rate does not only depend on the demand for the own occupation, but also on the demand for *skill-related* occupations.

Dependent variable-->	All unemployed		35 or younger		14 edu years or less	
	Random effects					
Unemployment growth						
Related openings (Da)	-0.0424***		-0.0292***		-0.0518***	
	(0.00710)		(0.00707)		(0.00986)	
Related openings (Dm)		-0.00052***		-0.000393***		-0.000564***
		(6.61e-05)		(6.58e-05)		(7.66e-05)
Own occupation openings	-0.103***	-0.0508**	-0.0908***	-0.0514**	-0.0945***	-0.0412*
	(0.0277)	(0.0234)	(0.0258)	(0.0213)	(0.0287)	(0.0242)
Real wage growth	-0.000853*	-0.00115**	0.00221	0.00214	-0.0155	-0.0150
	(0.000487)	(0.000518)	(0.00821)	(0.00821)	(0.0163)	(0.0161)
Constant	3.647***	3.790***	2.485***	2.609***	4.224***	4.342***
	(0.257)	(0.267)	(0.242)	(0.255)	(0.334)	(0.341)
Observations	1,933	1,933	1,914	1,914	1,437	1,437

Random effects. Standard errors clustered by occupation. 121 occupations and 16 States

Skill mismatch: employing from related occupations?

H2: The effect of the demand for *skill-related* occupations is stronger for younger workers.

Dependent variable-->	All unemployed		35 or younger		14 edu years or less	
	Random effects					
Unemployment growth						
Related openings (Da)	-0.0424***		-0.0292***		-0.0518***	
	(0.00710)		(0.00707)		(0.00986)	
Related openings (Dm)		-0.000522***		-0.00039***		-0.000564***
		(6.61e-05)		(6.58e-05)		(7.66e-05)
Own occupation openings	-0.103***	-0.0508**	-0.0908***	-0.0514**	-0.0945***	-0.0412*
	(0.0277)	(0.0234)	(0.0258)	(0.0213)	(0.0287)	(0.0242)
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Observations	1,933	1,933	1,914	1,914	1,437	1,437
Random effects. Standard errors clustered by occupation. 121 occupations and 16 States						

Skill mismatch: employing from related occupations?

H3: The effect of the demand for *skill-related* occupations is stronger for less-educated workers.

Dependent variable-->	All unemployed		35 or younger		14 edu years or less	
	Random effects					
Unemployment growth						
Related openings (Da)	-0.0424***		-0.0292***		-0.0518***	
	(0.00710)		(0.00707)		(0.00986)	
Related openings (Dm)		-0.000522***		-0.000393***		-0.00056***
		(6.61e-05)		(6.58e-05)		(7.66e-05)
Own occupation openings	-0.103***	-0.0508**	-0.0908***	-0.0514**	-0.0945***	-0.0412*
	(0.0277)	(0.0234)	(0.0258)	(0.0213)	(0.0287)	(0.0242)
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Observations	1,933	1,933	1,914	1,914	1,437	1,437

Random effects. Standard errors clustered by occupation. 121 occupations and 16 States

Conclusions

H1: Geographical distance between job openings and dormant skills reduces the job-finding frequency of the unemployed.

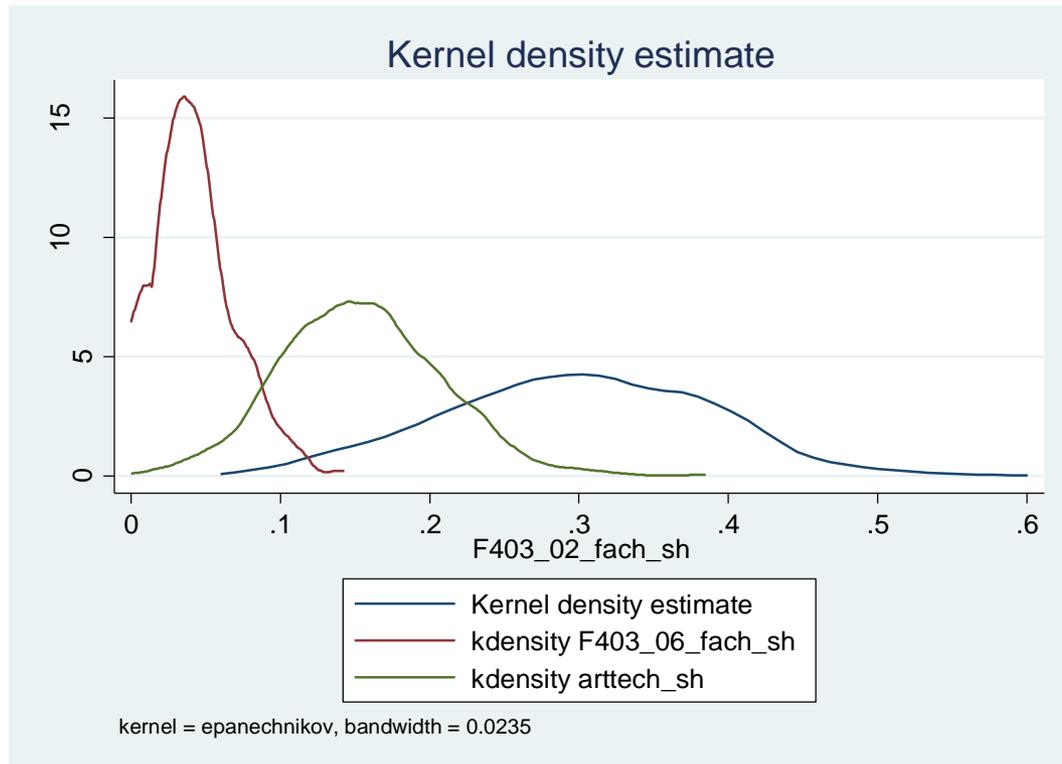
H2: Unemployed with past experience in occupations which are skill-related to the occupations in demand are more likely to become reemployed.

H3: Younger unemployed are less likely to become reemployed in skill-related occupations.

H4: Lower-skilled unemployed are more likely to switch to skill-related occupations.

- Policy implications

Appendix



Motivation

- How to counteract skill shortages and unemployment?
 - **Requalify unemployed**
 - “Das vorhandene Potenzial im Land sollte erst einmal genutzt werden. Wir können nicht zulassen, dass Menschen in Arbeitslosigkeit sind, nur weil ihre Talente nicht genutzt werden” (Frank-Jürgen Weise for Financial Times Deutschland, 02.08.2010)
 - “Priorität bei der Sicherung der Fachkräftebasis in Deutschland hat der Abbau der Arbeitslosigkeit” (BMAS, 2011, p. 18)

Dependent variable-->	Job-unemployment-job moves	All moves
Bordering region	-2.812*** (0.0279)	-2.832*** (0.0204)
Non-bordering region	-3.999*** (0.0297)	-4.125*** (0.0212)
Occupational distance	-5.716*** (0.124)	-6.230*** (0.0838)
Share of job openings in arrival region	0.0232*** (0.00174)	0.0268*** (0.00143)
Real wage level in arrival region	-0.0233*** (0.000642)	-0.00218*** (0.000475)
Constant	0.826*** (0.0535)	1.874*** (0.0473)
Inflated model (logit)		
Labour force	-0.0436*** (0.00203)	-0.0702*** (0.00100)
Constant	2.436*** (0.0294)	1.753*** (0.0178)
ln(alpha)	1.432*** (0.0484)	2.012*** (0.0157)
Observations	3,600,091	3,600,091

Robust standard errors in parentheses. Significant at the ***1%, **5%, *10% level.
Zero-inflated negative binomial model.