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IAB-DISCUSSION PAPER

Articles on labour market issues

7|2021 Industry interconnectedness and regional economic growth in Germany

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ISSN 2195-2663



Industry interconnectedness and regional economic growth in Germany

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Abstract

Urban systems, and regions more generally, are the epicenters of many of today's social issues. Yet they are also the global drivers of technological innovation and thus it is critical that we understand their vulnerabilities and what makes them resilient to different types of shocks. We take regions to be systems composed of internal networks of interdependent components. As the connectedness of those networks increases, it allows information and resources to move more rapidly within a region. Yet, it also increases the speed and efficiency at which the effects of shocks cascade through the system. Here we analyze regional networks of interdependent industries and how their structures relate to a region's vulnerability to shocks. Methodologically, we utilize a metric of economic connectedness, known as tightness, which attempts to quantify the ambiguous notion of a region's internal connectedness relative to other regions. Using industry employment, we calculate the economic tightness of German regions during the Great Recession, comparing it to each region's economic performance during the shock (2007–2009) and during recovery (2009–2011). We find that tightness is negatively correlated with changes in economic performance during the shock but positively correlated with performance during recovery. This suggests that regional economic planners face a tradeoff between being more productive or being more vulnerable to the next economic shock. Finally, we speculate on how these findings from the Great Recession may highlight potential implications of the ongoing COVID-19 pandemic and suggest future research that would compare outcomes of these two global shocks.

Zusammenfassung

Urbane Systeme und Regionen im Allgemeinen sind die Epizentren vieler heutiger sozialer Herausforderungen. Sie sind aber auch die globalen Treiber für technologische Innovationen. Daher ist es entscheidend, dass wir ihre Schwachstellen verstehen und wissen, was sie widerstandsfähig gegenüber verschiedenen Arten von Schocks macht. Wir betrachten Regionen als Systeme, die aus internen Netzwerken von voneinander abhängigen Komponenten bestehen. Wenn die Stärke dieser Netzwerke zunimmt, können Informationen und Ressourcen schneller innerhalb einer Region fließen. Dies kann jedoch auch die Geschwindigkeit und Stärke erhöhen, mit der sich die Auswirkungen von Schocks kaskadenartig durch das System ausbreiten können. Im vorliegenden Beitrag analysieren wir regionale Netzwerke von voneinander abhängigen Industrien und wie ihre Strukturen mit der Anfälligkeit einer Region für Schocks zusammenhängen. Methodisch verwenden wir eine Metrik der wirtschaftlichen Verflechtung (**“tightness“**), welche versucht, den vielschichtigen Begriff der internen Verflechtung einer Region relativ zu anderen Regionen zu quantifizieren. Anhand von Beschäftigungsdaten auf Branchenebene berechnen wir die wirtschaftliche Verflechtung der deutschen Regionen während der globalen Wirtschafts- und Finanzkrise und vergleichen sie mit der Entwicklung der Wirtschaftsleistung der einzelnen Regionen während des Schocks (2007–2009) und während der Erholung (2009–2011). Wir stellen fest, dass ein hohes Maß an **tightness** negativ mit der Wirtschaftsentwicklung während des Schocks, aber positiv mit der während der

Erholung korreliert ist. Dies deutet darauf hin, dass im Rahmen der kommunalen und regionalen Wirtschaftspolitik ein ausgewogener Mittelweg zwischen Produktivität und Resilienz gefunden werden sollte. Abschließend stellen wir erste Überlegungen dazu an, ob diese Erkenntnisse aus der Weltwirtschafts- und Finanzkrise auch auf die Implikationen der COVID-19-Pandemie übertragen werden können.

JEL classification

J40, R00, R11, R12, R58

Keywords

co-occurrence, COVID-19 pandemic, Great Recession, labor dynamics, industry structure, Panarchy, regional science, resilience, workforce, regional policy

Acknowledgements

We would like to thank Jeanette Carstensen sincerely for her effort and care in reformatting the original manuscript into this IAB Discussion Paper.

1 Introduction

Many of today's pressing social, economic, and ecological issues are focused in urban systems (Buhaug/Urdal 2013, President's Council of Advisors on Science and Technology 2016). They are epicenters of pollution, crime, inequality, and health problems. Yet urban systems, and the regions in which they are embedded, are also epicenters of creativity. They are the global drivers of innovation and productivity that fuel the global economy and the emergence of new technologies (Batty 2018, Vandecasteele et al. 2019). Being the focal point of so many social, ecological, and economic forces makes the management of regional development extremely difficult.

This complexity of regions has often been an obstacle to understanding and guiding regions and has resulted in policy interventions that have unintended and sometimes negative consequences. Fortunately, complexity science has begun to mature to the point that it is enabling researchers to move beyond lingering obstacles to a fundamental understanding of regional systems. Here we adopt the framework that regions are complex adaptive systems (Batty 2013, Lobo et al. 2020, McPhearson et al. 2016) and they exhibit a key attribute of such systems, namely internal networks of interdependent components (Barthelemy 2016, Shutters et al. 2018). As the internal connectedness of those networks increases, it can impact the ability for information and resources to move more rapidly and efficiently within a region. Yet, it also increases the speed and efficiency at which the effects of shocks cascade through the system.

In this study we focus on regional networks of interdependent economic components and how their structures relate to a region's vulnerability to shocks. To do so, we utilize a metric of economic connectedness, known as tightness, which attempts to quantify the rather ambiguous notion of a region's aggregate degree of internal interdependencies relative to other regions. Previously, this measure has only been analyzed in United States (U.S.) metropolitan areas (Shutters/Waters 2020, Shutters/Muneepeerakul/Lobo 2015, Waters/Shutters 2020) and has shown that regional tightness is positively correlated with more severe declines in economic performance following a shock (Shutters/Muneepeerakul/Lobo 2015). Thus, economic tightness is intimately linked to resilience and its importance lay in its potential ability to help anticipate impacts of system shocks, particularly economic shocks.

In our empirical analysis, we focus on Europe's biggest economy, Germany, for various reasons. Firstly, the country has emerged as one of the most resilient economies following thorough labor market and structural reforms at the beginning of the 21st century. Its gains in competitiveness vis-à-vis countries like France, Italy and Spain suggest that it was not the depreciation of the euro that served as the key driver of the German economy, but that it was Germany's high relative productivity gains (Dustmann et al. 2014). Our study focuses on another potential factor behind the economic success of Germany, the sectoral connectedness as a proxy for the strengths of regional supply chains as well as networks and thus important parameters for resilience. Secondly, regional disparities in Germany, even if decreasing between West and East Germany after the reunification of both parts in 1990, still persist (Heise/Porzio 2019) and were used as a justification for allocating more than 45 billion Euro for investment subsidies by the main instrument of German regional policy over the last 30 years (Bade/Alm/Weins 2020). Our analytical approach allows discussing if an increase in regional economic connectedness could lead to a decrease in regional

disparities. Finally, because German labor market regions (LMRs) cover all of Germany, they provide a valuable contrast to studies of U.S. metropolitan statistical areas, which cover only larger, urbanized areas of the U.S.

While others have used skills data to analyze resilience of a single German state (Otto/Nedelkoska/Neffke 2014), here we use industry employment data to calculate the economic tightness of all 141 German LMRs. We then compare that metric to each region's economic performance during the shock (2007–2009) and during the subsequent recovery (2009–2011). Using per worker GDP as a measure of economic performance, we find that tightness is negatively correlated with changes in economic performance during the shock but positively correlated with the same indicator during recovery. In other words, regions with tighter economies suffered more severely during a shock but had larger growth rates in the absence of a shock.

Based on the latest available employment statistics of the Federal Employment Agency (Bundesagentur für Arbeit), we also analyze the potential relationship between the economic tightness and the employment development of the German labor market regions in the first months of the COVID-19 pandemic. We observe that tightness seemed to play a significant role in regional economic development, even if the supply and demand triggered by the COVID-19 pandemic hit the German economy significantly more symmetrical than the 2007–2009 Great Recession as it impacted public life and therefore wide parts of the service and retail trade sector.

These overall results support the notion that economic interdependence can make regions more economically productive but also more precarious. This result concurs with the so-called Panarchy theory of complex adaptive systems (Holling 2001, Simmie/Martin 2010), which asserts that systems with higher internal connectedness are more efficient, but also more brittle and vulnerable to disruption. Thus, regional economic planners must navigate a tradeoff between being more productive or being more vulnerable to the next economic shock (Lobo et al. 2020, National Science Foundation Sustainable Urban Systems Subcommittee 2018).

2 Materials and Methods

2.1 Data and sources

Our geographical units of analysis are the 141 labor market regions (LMRs) of Germany as defined by Kosfeld and Werner (Kosfeld/Werner 2012). LMRs are composed of one or more of the 401 German counties ('Landkreise' and 'kreisfreie Städte') having close commuter links with each other and are considered to be essentially independent economic units. 'Landkreise' and 'kreisfreie Städte' comprise the NUTS3 units (Nomenclature des unités territoriales statistiques) of Germany and are roughly analogous to U.S. counties. Thus, we refer to these 401 smaller units as counties hereafter.

Industry employment data were supplied by Germany's Institute for Employment Research (IAB or Institut für Arbeitsmarkt- und Berufsforschung). Employment is analyzed at both 3-digit and 4-digit industry code aggregation levels.

To assess regional economic performance, we use regional gross domestic product (GDP) per capita. These estimates are taken from Federal Statistical Office (Destatis) data at the county level, aggregated to LMRs.

2.2 Quantifying interdependence, a pairwise measure of industry interdependence

To quantify an aggregate measure of interdependence characterizing each LMR's economy we first determine which industries are present or absent in each LMR. While a region may have employment in many industries, we use a threshold of relative abundance to determine presence or absence for our analysis. We apply the traditional metric of location quotient (LQ):

$$LQ_{i,m} = \frac{(e_{i,m}/\sum_i e_{i,m})}{(\sum_m e_{i,m}/\sum_m \sum_i e_{i,m})}. \quad (1)$$

where e is the employment in industry i in LMA m . Thus, LQ – also known as Revealed Comparative Advantage or RCA – is a region's proportion of employment in i compared to the national proportion. Following (Muneepeerakul et al. 2013) we take an industry i to be present in LMR m if $LQ_{i,m} \geq 1$ and absent if $LQ_{i,m} < 1$. This results in an LMR $\times$ industry matrix of presence-absence data.

This matrix is then used to calculate the probability that two industries i and j will co-occur in a random LMR. Using a variant of mutual information defined in (Muneepeerakul et al. 2013) we then quantify the interdependence x between any two industries i and j as

$$x_{i,j} = \frac{P[LQ_{i,m} > 1, LQ_{j,m} > 1]}{P[LQ_{i,m'} > 1]P[LQ_{j,m''} > 1]} - 1, \quad (2)$$

where m , m' , and m'' denote randomly selected LMRs. The result is an industry $\times$ industry matrix of interdependence values. Two industries that co-occur in LMRs more frequently than expected by chance will have a positive interdependence value, while two industries that co-occur less frequently than expected will have negative interdependence. We take this matrix of interdependence values to be the adjacency matrix of a complete weighted network describing how industries interact across Germany. Note that $x_{i,j} = x_{j,i}$ so that the adjacency matrix is symmetric and thus the network is undirected.

2.3 Quantifying tightness, a regional aggregate measure of industry interdependence

Using the methodology outlined in (Shutters/Muneepeerakul/Lobo 2015) we then calculate an aggregate measure of LMR interdependence, or tightness. This method dictates that we first assign an LMR-specific weight L to each pair of industries present in an LMR using the national-level pairwise interdependence x and weighting by the local proportions of employment in each industry:

$$L_{i,j,m} = \frac{(e_{i,m} + e_{j,m})x_{i,j}}{2 \sum_i e_{i,m}} \quad (3)$$

where i and j are industries that are both present in LMR m . We then average L across the total number of links in an LMR's industry subnetwork to produce an industry-based tightness metric:

$$T_m = \frac{2}{N_m(N_m-1)} \sum_{i < j}^{N_m} L_{i,j,m} \quad (4)$$

where i and j are both present in LMR m and N_m is the total number of industries present in LMR m . Note that averaging the weight L across all links in an LMR's subnetwork is equivalent to the generalized formula of network density (Shutters et al. 2018). Because tightness is a dimensionless measure, we normalize tightness values as a z-score with mean of 0 and standard deviation of 1.

2.4 Empirical treatments

The methodology described thus far applies to data for a specific year, industry aggregation level, and spatial unit. To examine the effects industry aggregation level, we calculate interdependence using employment data aggregated to both 3-digit ($N = 270$) and 4-digit ($N = 604$) industry levels. To better understand the effect of different regional definitions, we also calculated interdependence using both LMRs ($N = 141$) and counties ($N = 401$).

Thus, for a given year we construct four interdependence matrices using a 2×2 treatment approach. Finally, we replicate this 2×2 treatment using both 2007 and 2009 data for a total of eight interdependence matrices (industry level $\times$ area type $\times$ year). As our goal is to understand the relationship between economic tightness and regional vulnerability, years were chosen to correspond to points in time at the beginning of the 2007–2009 global recessions and at the beginning of the recovery period respectively.

3 Results and discussion

3.1 Industry interdependence

To calculate the economic tightness of LMRs we first calculated an interdependence value for all possible pairs of industries. The distributions of values using each skill type are shown as insets in Figure 1. Industry pairs that tend to appear together in the same LMRs will have higher interdependence than pairs that rarely occur together. Industry pairs that occur together less frequently than expected by chance will have a negative interdependence value.

Interdependence is a complex metric that captures elements of relatedness as defined by Frenken/Van Oort/Verburg (2007). These are industries sharing similar classification numbers in the German industry classification system. Thus, some pairs with high interdependence values are related in the Frenken sense, such as “Mining of hard coal” and “Manufacture of coke oven products”, which have the highest interdependence of any industry pair with $x = 22.5$ (using 2007 3-digit LMRs employment).

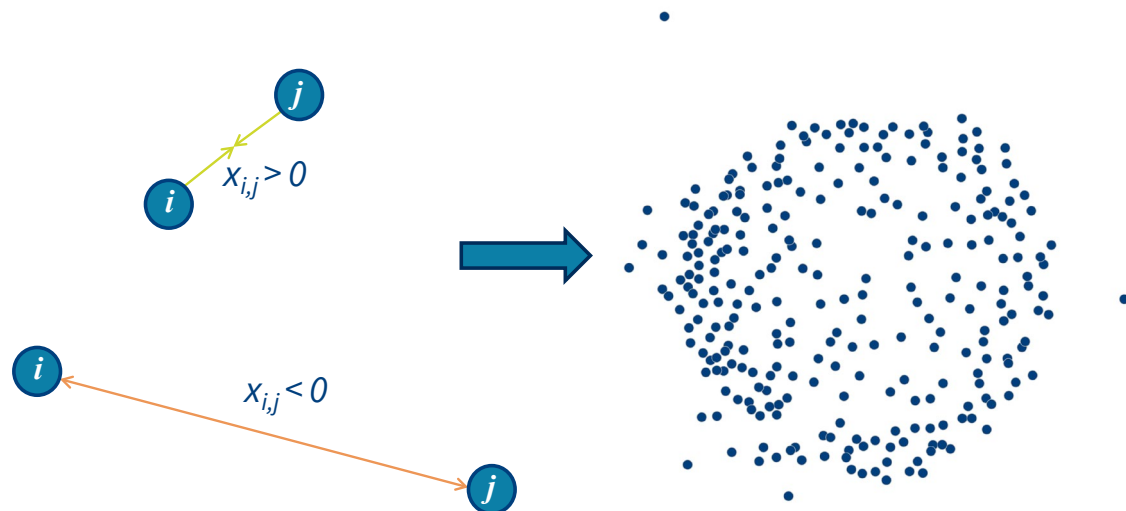
On the other hand, some highly interdependent pairs are seemingly unrelated in the Frenken sense, such as “Passenger air transport” and “Fund management activities” with $x = 19.1$. Thus, interdependence also captures aspects of relatedness described as so-called Marshallian externalities (O'Clery 2019, Essletzbichler 2015). These industries are related because they share common labor requirements, natural resource inputs, customers, etc.

3.2 The German industry network

Calculating interdependence between all industry pairs results in an industry $\times$ industry matrix of interdependence values. This matrix is taken to be the adjacency matrix of a complete and weighted national-level industry network, in which nodes are industries and weights are pair-wise interdependence values. The resulting network, shown in Figure 1, was created using MatLab's nonclassical multidimensional scaling function **mdscale**. Industry pairs that are highly interdependent tend to be near to each other in this network, while pairs with a low or negative interdependence tend to be farther apart.

In any LMR there exists only a subset of all possible industries. This industry portfolio represents an economic signature of the LMR and can be shown as a subnetwork of the national-level interdependence network. We take this subnetwork to be the “location” of a an LMR within a national map. Examples of two LMR subnetworks are shown in Figure 2. These examples demonstrate visually the concept of tightness as nodes are clustered more closely together in the LMR with higher tightness (Hamburg) and more dispersed in the region with lower tightness (Weißenburg-Gunzenhausen).

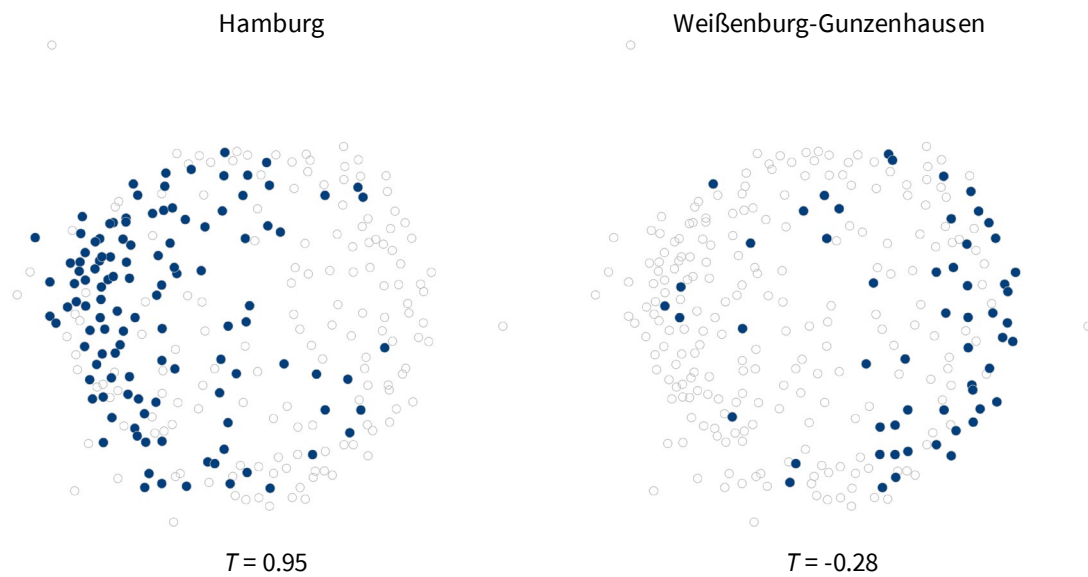
Figure 1: Constructing the 2007 German industry interdependence network



Industries that tend to co-occur together in regions have $x > 0$ and will tend to be closer to each other in the final network. Industries that rarely occur together will have $x \sim 0$ or $x < 0$ and will be pushed apart in the final network. Combining all network pairs results in the final German industry network (left). Each node represents a 3-digit industry and, while the placement of nodes is generally arbitrary, the distance between any two nodes is a function of an industry's interdependence x with all other industries. Links between nodes have been removed for visual clarity.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

Figure 2: Comparison of two German labor market regions (LMRs)



The subnetworks locations of Hamburg and Weißenburg-Gunzenhausen are shown within the 2007 German industry network of Figure 1. Qualitatively, the density of Hamburg nodes is higher than those of Weißenburg-Gunzenhausen. We quantify this visually intuitive difference in a regional tightness metric T . Links have been removed for clarity.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

The network shown in Figure 1 and Figure 2 was constructed with 2007 employment data at the 3-digit industry level for 141 LMRs. Though not shown, we also constructed networks for each of the other combinations of year, industry aggregation level, and spatial unit, as described in the empirical treatments of the methods section. Thus, we ultimately created eight national industry interdependence networks (2007/2009, 3-/4-digit industry code, 141 LMRs/401 counties).

3.3 LMRs and tightness

Applying industry interdependence values to the eight networks described in the previous section, we then located the subnetwork of every German spatial unit within the corresponding network and calculated its tightness T . A list of LMRs having highest and lowest 2007 tightness values using 3-digit industry data are presented in Table 1.

The distribution of tightness values across regions is highly skewed with only four regions being more than two standard deviations from the mean T value. The highest level of interconnectedness in the LMR Wolfsburg is due to the high importance of the automotive industry there, which is much more pronounced in this region than in any other German LMR. About one third of all employees in and around Wolfsburg work in the automotive and automotive supplier industry in 2008. Employee leasing, accounting for 6 percent of all employees, may also play a role here.

As the capital region, Berlin is strongly characterized by employees in political interest groups, but also in information technology services, universities and research institutes, as well as credit institutes. The Frankfurt am Main region is Germany's central banking district. In addition, many employees in aviation, management consultancies are concentrated here. The regions of Frankfurt an der Oder and Cottbus are characterized by larger shares of employees in steel production and lignite mining, respectively, and power generation (coal-fired power generation). Both sectors are

characterized by significant direct and indirect employment effects on other economic sectors in the region but also beyond.

LMRs at the lower end of the tightness table, on the other hand, are predominantly characterized by public-sector industries such as hospitals, geriatric care, social services, public administration, and retail. In addition, the five regions feature singular concentrated manufacturing industries that appear to have fewer links to other industries, such as furniture manufacturing or plastic goods manufacturing.

Table 1: Highest and lowest ranked Labor Market Regions based on 2007 tightness values T

Rank	Labor Market Region (LMR)	T ¹⁾
1	Wolfsburg (007)	8.27
2	Berlin (109)	5.10
3	Frankfurt am Main (043)	2.68
4	Frankfurt an der Oder (110)	2.05
5	Cottbus (118)	1.68
137	Trier (054)	-0.61
138	Koblenz (049)	-0.61
139	Kleve (027)	-0.62
140	Minden (036)	-0.64
141	Passau (086)	-0.64

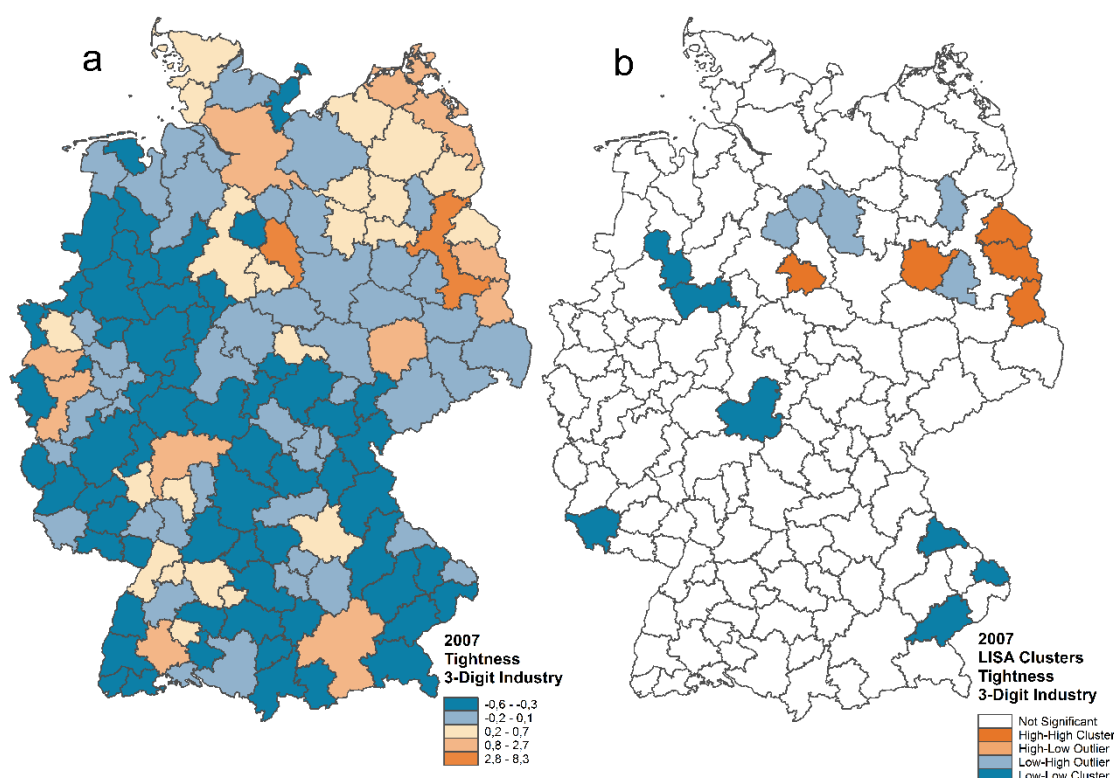
1) T presented as the normalized z-score of raw tightness values.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

3.4 Spatial distribution and autocorrelation of tightness

Examining the spatial distribution of 2007 tightness we find that LMRs with high tightness are more prevalent in the former East (Figure 3a), while LMRs with low tightness are prevalent in the west and south of Germany. To identify statistically significant clusters of LMRs with high and low tightness, we applied Anselin's Local Moran's I (Anselin 1995), a metric of spatial association. Spatial clusters are neighboring regions with tightness values more similar to one another than would be anticipated under spatial randomness. The areas of interest are "high-high" and "low-low", which are statistically significant spatial clusters of high and low tightness respectively. Results presented in Figure 3b confirm that a cluster of labor markets with high tightness existed in 2007 in eastern Germany.

Figure 3: Spatial distribution and autocorrelation of LMR tightness values, 2007



(a) Tightness and (b) Anselin's Local Moran's I (LISA) shown here were calculated at the 3-digit industry code level. Clusters shown in (b) are significant at a confidence level of 0.05 using k -nearest neighbors = 4.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

3.5 Tightness, productivity, and economic shocks

Having calculated each area's tightness under each of the eight empirical treatments, we then correlated those values with the area's economic performance. We take 2-year change in per capita GDP as our measure of economic performance, calculating it for the periods 2007–2009 and 2009–2011. Correlations for the eight empirical case are presented in Table 2. In all cases, tightness is negatively correlated with changes in GDP during the shock and positively correlated with changes in GDP during recovery. Furthermore, we find that correlations are stronger when using 3-digit industry aggregations versus 4-digit industry aggregations, and they are stronger when using labor market regions versus counties.

This suggests that LMRs are more functional regarding regional economic attributes than counties. It also suggests that 4-digit industry aggregations may be too sparse to meaningfully inform our metric of connectedness. Thus, we focus our discussion of this work primarily on metrics calculated using LMRs and 3-digit industry data.

Overall, our finding that regions with higher tightness had larger percentage drops in productivity following a shock is consistent with the Panarchy theory of resilience, which asserts that as systems increase in connectedness, they become more brittle and fragile (Gunderson/Holling 2002). This finding also concurs with an analogous study of the effects of the 2007–2009 recession on U.S. metropolitan statistical areas (Shutters/Muneepeerakul/Lobo 2015).

Table 2: Correlation between tightness T and change in GDP per worker over the subsequent two years

	Labor Market Regions		Counties	
Industry Level	3-digit	4-digit	3-digit	4-digit
T (2007) v. GDP change, 2007-2009	-0.22 ***	-0.19 **	-0.20 ***	-0.09 n/s
T (2009) v. GDP change, 2009-2011	0.34 ***	0.25 **	0.27 ***	0.14 **v

* significant at $p < 0.05$; ** significant at $p < 0.01$; *** significant at $p < 0.001$; n/s = not significant.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

However, the fact that regions with high tightness performed better during the recovery phase should be interpreted with caution. The recovery period following a shock may not be comparable to periods without shocks and so we cannot conclude that tighter economies are, in general, more productive.

3.6 A first comparison of tightness effects and COVID-19 implications

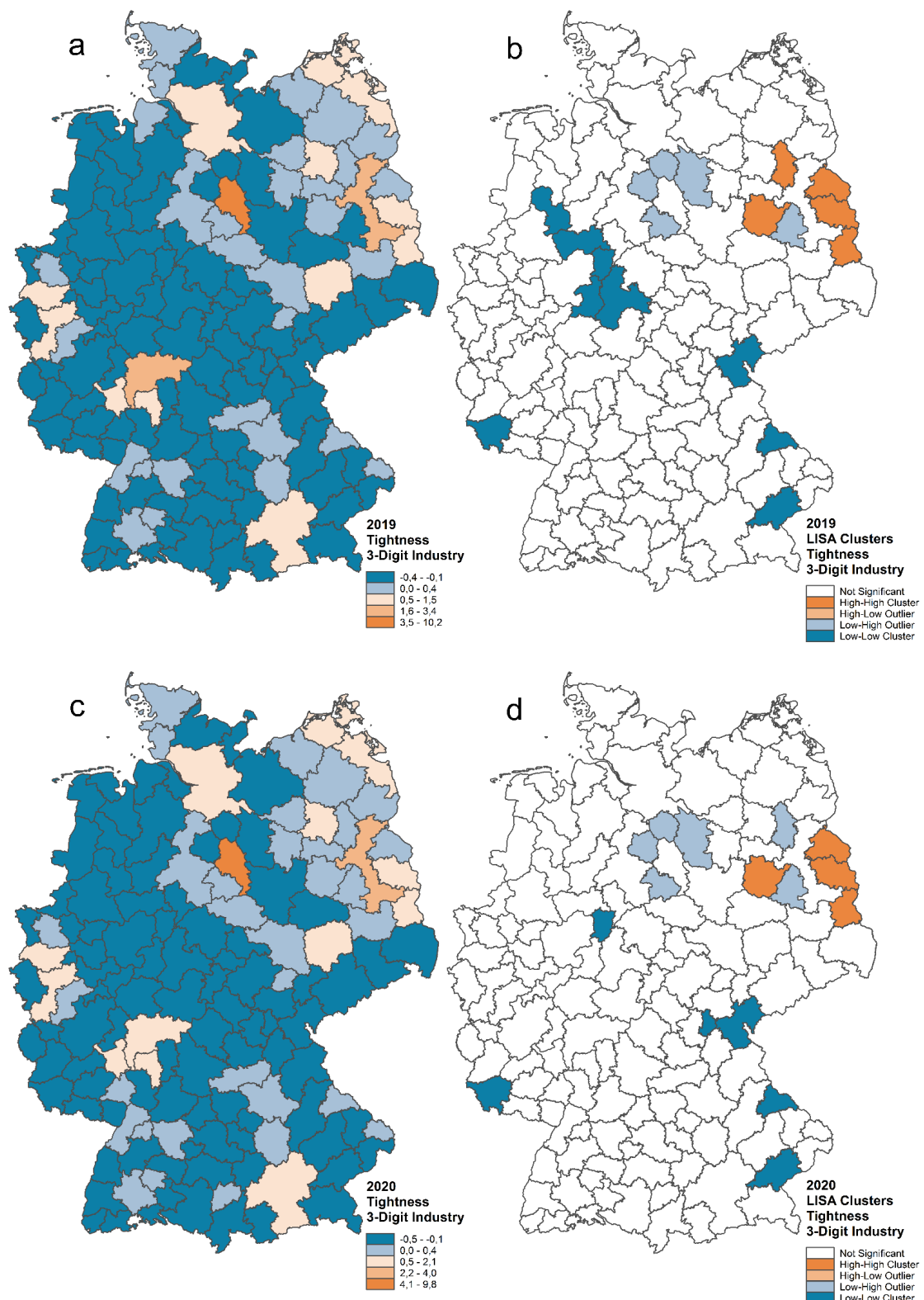
At first glance, the impact of the COVID-19 pandemic on the German economy was quite comparable with the Great Recession: the GDP in 2020 declined by 5.0 per cent, which was very similar to 2009 (-5.6 %). But does this correspond with a similar association between tightness and regional economic performance during these two shocks?

Having demonstrated that tightness in 2007 was correlated with worse performance during the subsequent shock, we calculate tightness during the first months of the COVID-19 pandemic using the latest years of available employment data. The spatial distribution of tightness for 2019 and 2020 is presented in Figure 4.

Our analysis suggests that the effect of the COVID-19 pandemic on regional employment (regional GDP data were not yet available at the time of this study) was very different from the Great Recession. Comparing September to December 2008 employment development with March to June 2020 regional employment declines are distributed more evenly across counties in the COVID-19 crisis than in the Great Recession. Also short-time work – an instrument of German labor market policy that subsidizes wages for staff that had otherwise to be laid-off due to the economic conditions – varies much more evenly across regions in the pandemic than in the Great Recession.

Figure 4 also clearly indicates that tightness during the COVID-19 pandemic was not as negatively correlated with changes in economic performance as during the Great Recession. The crucial factor for this is that the COVID-19 pandemic was more symmetric in terms of sectors being affected significantly negative by the supply shock and contraction of demand whereas the Great Recession primarily hit export-oriented manufacturing industries and only parts of the services sector (especially financial services). Once more data becomes available on the regional economic performance during and after the Covid-19 pandemic, our results should be set into a broader perspective to cover the impact of tightness on GDP and employment both during a shock and the recovery.

Figure 4: Current spatial distribution and autocorrelation of LMR tightness values, 2019 and 2020



Tightness (a and c) and Anselin's Local Moran's I (LISA) (b and d) were calculated at the 3-digit industry code level. Clusters shown in (b) are significant at a confidence level of 0.05 using k-nearest neighbors = 4. Between 2007 and 2019 the cluster of high-high tightness areas in eastern Germany grew in size while the clusters of low-low tightness in central Germany also grew. Thus, the most recent employment data indicates a growing spatial segregation within Germany in terms of economic tightness.

Source: Employment statistics of the Federal Employment Agency, Germany. Own calculations.

4 Conclusion

In this study, we have used a novel metric to assess the level of internal connectedness of German regions and to assess how that measure relates to economic performance during and after the shock of the Great Recession. Ours is the first study to apply this metric to a country other than the U.S.

Like U.S. results, we found that regions with more tightly connected industry structures had larger drops in per capita GDP during the initial shock and larger growth during the recovery. This suggests a tradeoff faced by regional economic policy makers between higher productivity and higher resilience.

However, our study examined a very specific shock in terms of type and scope and we are careful not to generalize our results to all shocks a region might face. As shown in section 3.6, pandemics and other biological threats seem to yield – given a specific industry interconnectedness – different economic outcomes, this caution seems to be justified. For other types of shock, such as trade wars, natural disasters like climate shocks/extreme weather events, or cyber-attacks, our metric of tightness may also suggest different implications for economic performance.

Further research is needed to clarify the role of industry interconnectedness under different types of shock. On a national level, our metric could be used to assess if a certain degree of tightness might increase the resilience or precariousness of supply chains during disruptions, and thus promote economic and national security as well as the capacity to respond to external shocks.

References

- Anselin, Luc (1995): Local Indicators of Spatial Association–LISA. In: *Geographical Analysis*, 27 (2), pp. 93–115.
- Bade, Franz-Josef; Alm, Bastian; Weins, Sebastian (2020): Employment effects of investment subsidies by German Regional Policy. In: *EconStor Preprints* 218843, ZBW – Leibniz Information Centre for Economics.
- Barthelemy, Marc (2016): *The Structure and Dynamics of Cities: Urban Data Analysis and Theoretical Modeling*. Cambridge: Cambridge University Press.
- Batty, Michael (2013): *The New Science of Cities*. Cambridge, Massachusetts: MIT Press.
- Batty, Michael (2018): *Inventing Future Cities*. Cambridge, Massachusetts: MIT Press.
- Buhaug, Halvard; Urdal, Henrik (2013): An urbanization bomb? Population growth and social disorder in cities. In: *Global Environmental Change*, 23 (1), pp. 1–10.
- Dustmann, Christian; Fitzenberger, Bernd; Schönberg, Uta; Spitz-Oener, Alexandra (2014): From Sick Man of Europe to Economic Superstar: Germany's Resurgent Economy. In: *Journal of Economic Perspectives*, 28 (1), pp. 167–188.
- Essletzbichler, Jürgen (2015): Relatedness, Industrial Branching and Technological Cohesion in US Metropolitan Areas. In: *Regional Studies*, 49 (5), pp. 752–766.
- Frenken, Koen; Van Oort, Frank; Verburg, Thijs (2007): Related Variety, Unrelated Variety and Regional Economic Growth. In: *Regional Studies*, 41 (5), pp. 685–697.
- Gunderson, Lance H.; Holling, C. S. (eds.) (2002): *Panarchy: understanding transformations in human and natural systems*. Washington, DC: Island Press.
- Heise, Sebastian; Porzio, Tommaso (2019): Spatial Wage Gaps in Frictional Labor Markets. In: *Opportunity and Inclusive Growth Institute, Working Papers No. 29*, Federal Reserve Bank of Minneapolis.
- Holling, C. S. (2001): Understanding the Complexity of Economic, Ecological, and Social Systems. In: *Ecosystems*, 4 (5), pp. 390–405.
- Kosfeld, Reinhold; Werner, Alexander (2012): German Labour Markets—New Delineation after the Reforms of German District Boundaries 2007–2011. In: *Raumforschung und Raumordnung (RuR)*, 70 (1), pp. 49–65.
- Lobo, Jose; Alberti, Marina; Allen-Dumas, Melissa; Arcaute, Elsa; Barthelemy, Marc; Bojorquez Tapia, Luis A.; Brail, Shauna; Bettencourt, Luis; Beukes, Anni; Chen, Wei-Qiang; Florida, Richard; Gonzalez, Marta; Grimm, Nancy; Hamilton, Marcus; Kempes, Chris; Kontokosta, Constantine E.; Mellander, Charlotta; Neal, Zachary P.; Ortman, Scott; Pfeiffer, Deirdre; Price, Michael; Revi, Aromar; Rozenblat, Céline; Rybski, Diego; Siemiatycki, Matthew; Shutters, Shade T.; Smith, Michael E.; Stokes, Eleanor C.; Strumsky, Deborah; West, Geoffrey; White, Devin; Wu, Jingle; Chuqiao Yang, Vicky; York, Abigail; Youn, Hyejin (2020): Urban Science: Integrated Theory from the First Cities to Sustainable Metropolises. In: *SSRN Electronic Journal*.

- McPhearson, Timon; Haase, Dagmar; Kabisch, Nadja; Gren, Åsa (2016): Advancing understanding of the complex nature of urban systems. In: *Ecological Indicators*, 70 (November 2016), pp. 566–573.
- Muneepeerakul, Rachata; Lobo, José; Shutters, Shade T.; Gómez-Liévano, Andrés; Qubbaj, Murad R. (2013): Urban Economies and Occupation Space: Can They Get "There" from "Here"? In: *Plos One*, 8 (9), e73676.
- National Science Foundation Sustainable Urban Systems Subcommittee (2018): Sustainable Urban Systems: Articulating a Long-Term Convergence Research Agenda. A Report from the NSF Advisory Committee for Environmental Research and Education. Washington, DC: U.S. National Science Foundation.
- O'Clery, Neave; Heroy, Samuel; Hulot, Francois; Beguerisse-Diaz, Mariano (2019): Unravelling the forces underlying urban industrial agglomeration. In: *arXiv*, 1903.09279v2.
- Otto, Anne; Nedelkoska, Ljubica; Neffke, Frank (2014): Skill-relatedness und Resilienz: Fallbeispiel Saarland. In: *Raumforschung und Raumordnung*, 72 (2), pp. 133–151.
- President's Council of Advisors on Science and Technology (2016): Technology and the Future of Cities: Report to the President. Washington, DC: Executive Office of the President of the United States, President's Council of Advisors on Science and Technology.
- Shutters, Shade T.; Lobo, Jose; Strumsky, Deborah; Muneepeerakul, Rachata; Mellander, Charlotta; Brachert, Mathias; Farinha Fernandes, Teresa; Bettencourt, L. M. A. (2018): The relationship between density and scale in information networks: The case of urban occupational networks. In: *Plos One*, 15 (5), e0196915.
- Shutters, Shade T., Rachata Muneepeerakul, and José Lobo (2015): Quantifying urban economic resilience through labour force interdependence. In: *Palgrave Communications*, 1 (201510), pp. 1–7.
- Shutters, Shade T.; Waters, Keith (2020). Inferring Networks of Interdependent Labor Skills to Illuminate Urban Economic Structure. In: *Entropy*, 22 (10), p. 1078.
- Simmie, James; Martin, Ron (2010): The economic resilience of regions: towards an evolutionary approach. *Cambridge Journal of Regions, Economy and Society*, 3(1):27-43.
- Vandecasteele, Ine; Baranzelli, Claudia; Siragusa, Alice; Aurambout, Jean Philippe (eds.) (2019): The Future of Cities—Opportunities, challenges and the way forward. Luxembourg: Publications Office of the European Union.
- Waters, Keith; Shutters, Shade T. (2020): Industrial structure and a tradeoff between productivity and resilience. In: *SSRN Electronic Journal*. URL: <https://doi.org/10.2139/ssrn.3702545>.

Imprint

IAB-Discussion Paper 7|2021

Date of publication

28 April 2021

Publisher

Institute for Employment Research
of the Federal Employment Agency
Regensburger Str. 104
90478 Nürnberg Germany

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ISSN

2195-2663

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