

A Service of



Leibniz-Informationszentrum Wirtschaft Leibniz Information Centre for Economics

Braun, Stefanie; Lee, Gabriel S.

Working Paper The prices of residential land in German counties

BGPE Discussion Paper, No. 194

Provided in Cooperation with:

Friedrich-Alexander University of Erlangen-Nuremberg (FAU), Bavarian Graduate Program in Economics (BGPE)

Suggested Citation: Braun, Stefanie; Lee, Gabriel S. (2020) : The prices of residential land in German counties, BGPE Discussion Paper, No. 194, Friedrich-Alexander-Universität Erlangen-Nürnberg, Erlangen und Nürnberg

This Version is available at: https://hdl.handle.net/10419/237985

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.

Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.



WWW.ECONSTOR.EU



BGPE Discussion Paper

No. 194

The Prices of Residential Land in German Counties

Stefanie Braun Gabriel S. Lee

April 2020

ISSN 1863-5733

Editor: Prof. Regina T. Riphahn, Ph.D. Friedrich-Alexander-Universität Erlangen-Nürnberg © Stefanie Braun, Gabriel S. Lee

The Prices of Residential Land in German Counties^{*}

Stefanie Braun University of Regensburg Universtitaetsstrasse 31 93053 Regensburg, Germany

Gabriel S. Lee University of Regensburg Universitiaetsstrasse 31 93053 Regensburg, Germany And Institute for Advanced Studies Josefstaetterstr. 39 1080 Vienna, Austria <u>Contact Information:</u> Braun: + 49 941 943 5086; E-mail: stefanie.braun@ur.de Lee: + 49 941 943 5060; E-mail: gabriel.lee@ur.de

April 8, 2020

Abstract

We estimate changes in the value and price of residential land for 379 German counties ("Landkreise") from 2014 to 2017 using a total of 42,685 observations. We use the two-step residual method that decomposes the value of a home into the value of the structure and land value. Despite the short time series, we show that the price of residential land has become relatively more expensive in the majority of German counties. More specifically, we show that the cumulative change in land values varies between 9% and 171% (excluding Berlin with exorbitant high increases of 668% and Saarland with a decrease of 18%) from 2014 to 2017. On average, the home values increased by 30% and 15% for West and East Germany respectively, whereas the land values increased by 108% and 91% for West and East Germany. Our findings imply that cycles in the German land values are more likely to affect the evolution of house prices more in the future than they did in the past. Moreover, our estimated land prices vary significantly to the current land price valuation by one of the German state governments (Hessen).

- JEL Classification: R0; R11; R14; R21; R31
- *Keywords:* German Land prices; Land values; German Housing prices; Housing values; Construction costs; Replacement costs

^{*}We would like to thank Morris Davis, Fabian Kindermann, and the seminar participants at the University of Regensburg for constructive comments. We also thank Vdp Research and BKI for providing house prices and construction cost data. The usual disclaimer applies.

1 Introduction

Figure 1 shows the housing wealth (including both structures and land) of households and housing prices for Germany from 2003-2017.¹ Three salient facts in Figure 1 are first, housing accounts for 62% of national wealth in Germany at the end of 2017. Second, the housing prices have been sharply increasing in the last ten years. Lastly, the fraction of land in total wealth has been increasing over the last six years, but the fraction of dwellings' value in total wealth has been steadily decreasing. Moreover, boom and bust cycles in the housing sector occurred antecedent to several past great recessions and especially before the most recent financial crisis of 2008-2010. These facts suggest that house price changes may have significant effects on the German macroeconomy. As in Davis and Heathcote (2007) and Davis and Palumbo (2008) for the U.S., we also argue that the movements in the land - and structure values are crucial for understanding the development of German housing markets.

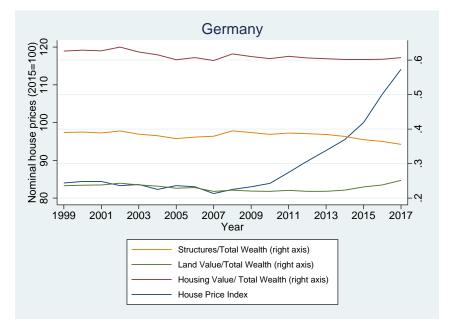


Figure 1: Housing Prices, Housing Wealth for Germany: 1999-2017 Annual

The objective of this paper is to estimate changes in the value and price of residential land for the 402 German counties ("Landkreise")² from 2012 to 2017. The framework of our approach is that of Davis and Heathcote (2007), who decompose the value of a home into the value of the structure and land value on the aggregate level for the U.S., and Davis and Palumbo (2008), who estimate the land prices for 46 large U.S. metropolitan areas. Previous works on German land prices such as Kolbe, Schulz, Wersing, and Werwatz (2012, 2015, and 2019) use land value assessment methods using transaction data from local surveyor commissions and focus only on the city of Berlin. Kolbe, Schulz, Wersing, and Werwatz (2012, 2015, 2019) generate highly correlated land value estimates to expert based assessments. However, most of their methods

¹From the OECD national accounts data set, we construct the total national wealth as the sum of fixed nonfinancial assets and land, and the net financial wealth. For the housing wealth, we use the total values of dwellings and land.

²Official administrative boundaries as of 2014.

1 INTRODUCTION

require high computational power suggesting that their methods are mostly applicable on a very local level. This paper, however, estimates German land prices on county levels using housing price indices.

Our motivation in estimating the German land prices on county levels using Davis and Palumbo (2008) framework is to provide a systematic method that can address and help the following issues in German housing markets. Germany uses two land value measurements. The first measurement bases on the statistics of purchased values of building land that reports purchase/selling prices and other characteristics of undeveloped land.³ The second measurement uses the land reference values from the expert committees for land values.⁴

Consequently, existing large-scale measurements of land values in Germany only refer to undeveloped land or are estimated by local surveyor commissions in German counties and countyfree cities with some inconsistent standards and time intervals. Hence, these measurements are not comparable for all single-/double family housing units throughout Germany. Subsequently, our residual method provides an unbiased estimate of land prices for single-/double family housing on the county level for the whole of Germany. Moreover, our land price estimates could further provide an alternate view for real estate market participants in evaluating the economic feasibility of their projects, and help policymakers in addressing some of the housing regulation laws. In particular, our estimates help to implement the current German property tax reform that requires a frequent, cost-effective land price valuation methodology.⁵

One of the main results from both Davis and Heathcote (2007) and Davis and Palumbo (2008) is that the prices of both components of the housing bundle evolve quite differently. Consequently, they shed light on the importance of distinguishing between the values of construction costs and residential land when analyzing home prices. Without the decomposition of housing prices, Davis and Palumbo (2008) argue that regression results from housing market-related variables on the fundamentals in determining housing prices can lead to misguiding conclusions. Furthermore, both Davis and Heathcote (2007)) and Davis and Palumbo (2008) show that land prices appear to be three times more volatile at business cycle frequencies than the value of the structural component. Moreover, the average share of a home's value attributed to residential land increased significantly between 1984 and 2004, and while the inflation-adjusted price of land increased by more than 400%, the price of housing structures appreciated by only 33%. Consequently, to estimate and identify land prices in a systematic method is crucial in understanding housing prices.

As with Davis and Palumbo (2008), we also show that despite the short time series, the price

 $^{^{3}}$ These statistics are available at the regional statistics website from the statistical offices of the German federation and countries under statistic number 61511.

⁴Two critiques on the expert valuation method. First, the method is extremely ad hoc. A second critique is that the valuation occurs irregularly: the valuations are decided anually or biennially either on December 31st of the previous year or January 1st of the current year. (See Paragraph 196 BauGB).

⁵The calculation of the property tax is based on decades old property values. For West German countries, they are based on land values from 1964, and for East German countries on values of the year 1935. As land values have developed quite differently since these years, the deadline for a new regulation is set at December 31, 2019, in order to eliminate unequal tax treatment. If a new regulation is introduced by the end of 2019, then the old rules can still be applied until end of 2024. Starting 2025 the new legal regulation must then be used. (Source: Federal Ministry of Finance).

of residential land has become relatively more expensive in the majority of German counties. The counties around urban centers such as Munich, Stuttgart, Berlin, Hamburg, Dresden and the cities in the Ruhr area experienced the highest land price increases. But in general, an upward shift in home values, land values, and the share of residential land occured in almost every state. The most significant differences in the changes in land prices are between the new and old federal states of Germany. Although the differences in the average land's share of home value increased between East and West, the absolute differences in home values, the value of residential land, and replacement costs did not decrease. More specifically, the cumulative change in land values from 2014 to 2017 varies between 9% and 171% (excluding Berlin with exorbitant high increases of 668% and Saarland with a decrease of 18%). On average, the home values increased by 30% and 15% for West and East Germany. Our findings imply that cycles in the German land values are likely to influence the evolution of house prices more in the future than they did in the past.

The next section briefly overviews the literature. We then outline the methodology in Section 3 as well as the detailed description of Davis and Heathcote (2007) and Davis and Palumbo (2008) version for Germany. Our empirical results and discussions are in Section 4. Section 5 concludes followed by the Appendix.

2 Literature Review

The literature on the estimation of land prices can broadly be attributed to two strands of literature. First, the residual method by Davis and Heathcote (2007) and Davis and Palumbo (2008) measures the price of land as the difference between housing value and the replacement cost of the structure. The second line of research follows spatial transaction based approaches that use data on sales of vacant land as in, for example, Albouy, Ehrlich, and Shin (2018), Nichols, Oliner, and Mulhall (2013) and Haughwout, Orr, and Bedoll (2008).

Davis and Heathcote (2007) and Davis and Palumbo (2008) decompose the value of a home into the value of the structure and land value, both on the aggregate level for the U.S. as well as for 46 large U.S. metropolitan areas. The two-step procedure calculates estimates of replacement costs of residential structures in a given base year and builds aggregate structure share for a given regional unit in that year. In the second step, the benchmark estimates for structures share and replacement costs are extrapolated, and the growth in land prices over the time series is derived. They show the importance of distinguishing between values of construction costs and residential land rather than using exclusively home prices in addressing macroeconomic and housing market implications.

Recent work by Davis, Oliner, Pinto, and Bokka (2017) extends the two-step residual method by focusing on new homes using proprietary data on zip-code level for the Washington DC metro area. By doing so, Davis et al. (2017) avoid estimating the depreciation of structure costs, and hence reducing the amount of estimation error. Using appraisal data for more than 900 counties, 8,000 ZIP codes and 11,000 census tracts in the United States, Davis, Larson, Oliner, and Shui (2019) further exclude homes with an effective age of more than ten years given the predictions of a calibrated option model of teardowns to ensure that market values are represented by estimated replacement cost. Davis et al. (2019) also account for the size of a lot when estimating land values to standardize land values to a per acre basis. They then apply spatial interpolation methods to estimate land prices also for housing units without appraisal reports.

The spatial transaction-based approaches, for example, by the aforementioned authors, use observed market price data on transactions of vacant or almost vacant land (for the U.S. mainly from the CoStar Group Inc.). They estimate land prices for either only one, several, or close to the universe of metropolitan areas in the U.S. Key assumption in these methods is that the closer land is to the city center, the more valuable it is. Typical findings of this literature are that land prices in the United States tend to be more volatile than structure prices, which likely arises from the fact that the supply of structures is much more elastic than the supply of land (e.g., Gyourko & Saiz, 2006). This finding supports the so-called land leverage hypothesis, which implies that home prices are more volatile in areas with a relatively large share of land value in housing value (e.g., Bostic, Longhofer, & Redfearn, 2007).

Kolbe et al. (2012, 2015, 2019) focus on German land prices. In particular, they present different land value assessment methods using transaction data from local surveyor commissions from Berlin, Germany. They use a semiparametric method to construct location values. First, they separate the house value into structure and land components. Second, using different nonparametric regression methods such as the adaptive weight smoothing approach, they separate the first stage residual into the location value and a noise term. In contrast to kernel regressions, for example, adaptive weight smoothing allows the location value surface to be determined entirely by the data. They generate highly correlated estimates with expert-based assessments.

3 Methods and Data

We follow Davis and Heathcote (2007) and Davis and Palumbo (2008) in constructing our land price index. For the expositional purpose, we briefly outline the method below. We then describe and present in detail how we merge different sources of data to compute annual time series estimates for the average value of land in German counties and county-free cities.

We start with the nominal market value of residential land at date t:

$$p_t^l l_t = p_t^h h_t - p_t^s s_t. (1)$$

We denote p_t^l , p_t^h and p_t^s as the quality-adjusted prices per unit of land, houses and structures; and l_t, h_t and s_t represent the quality-adjusted quantities. Hence, $p_t^h h_t$ is the market value of the housing stock and $p_t^s s_t$ the replacement cost of stock of residential structures and "land" refers to everything that increases home value by more than the replacement cost.

Assuming that houses in a set observed over two successive periods do not change (i.e., structures and land are constant), we can express the growth rate in house prices using the

3 METHODS AND DATA

decomposition from Equation (1) as

$$\frac{p_{t+1}^{h}h_{t}}{p_{t}^{h}h_{t}} = \frac{p_{t+1}^{s}s_{t} + p_{t+1}^{l}l_{t}}{p_{t}^{h}h_{t}} = \frac{p_{t}^{s}s_{t}(p_{t+1}^{s}/p_{t}^{s}) + p_{t}^{l}l_{t}(p_{t+1}^{l}/p_{t}^{l})}{p_{t}^{h}h_{t}} = w_{t}^{s}\frac{p_{t+1}^{s}}{p_{t}^{s}} + w_{t}^{l}\frac{p_{t+1}^{l}}{p_{t}^{l}},$$
(2)

where $w_t^s = \frac{p_t^s s_t}{p_t^h h_t}$ and $w_t^l = \frac{p_t^l l_t}{p_t^h h_t}$ are the shares of home value that are accounted for by replacement cost of structures and market value of land in period t, respectively. As these shares sum to one, we can rearrange Equation (2) to express the percentage changes of (constant-quality) land prices for a certain county i in period t as follows:

$$g_{it+1}^{lp} = \frac{1}{w_t^l} [g_{it+1}^{hp} - (1 - w_t^l) g_{it+1}^{cc}].$$
(3)

where, g_{it+1}^{lp} is the value-weighted average growth rate of residential land containing the existing stock of homes in a county *i* between period *t* and *t* + 1. It is important to note that the index is simply the growth rate of the price of all attributes (including location and amenities) that make the homes in a county more expensive than the replacement costs of their structures. It is, however, not a "euro-per-square-meter" concept. To estimate changes in residential land prices for German counties, we apply a two-step procedure similar to Davis and Palumbo (2008) to obtain time series for w_t^l and w_t^s . The next section describes in detail the procedure and data for Germany. However, in short, the first step involves estimating structures share at a benchmark date using real estate data for Germany. In the second step, given these estimated structure shares at a benchmark date, growth in construction cost and house prices, we apply the following dynamic equation to construct a time series of structure shares.:

$$w_{it}^{s} = w_{it-1}^{s} \left(\frac{g_{it}^{cc}}{g_{it}^{hp}}\right) \left(\frac{h_{it-1}}{h_{it}}\right) + \theta_{it} \frac{\Delta h_{it}}{h_{it}}$$
(4)

where, $g_{it}^{cc} = \frac{p_{t+1}^s}{p_t^s}$ is the growth rates in construction cost, $g_{it}^{hp} = \frac{p_{t+1}^h}{p_t^h}$ is the growth rate of house prices, $\frac{h_{it-1}}{h_{it}}$ is a proxy for the growth rate of real housing stock, and θ_t is a value for the structure intensity of nominal net new housing.

Some of the implications of Equation (4) are worthwhile noting. First, when there is no growth in the housing stock $(h_t = h_{t-1})$, then the structures share in period t equals the structures share of period t-1 adjusted for growth in construction cost relative to house prices. Second, for growing cities $(h_t > h_{t-1})$, the growth in construction cost relative to house prices is relatively less important in determining structures share of the next period. Instead, the structures share of new housing plays a larger role.

3.1 Benchmark Estimates for the Replacement Cost of Residential Structures: German Case

In this section, we describe our approach in computing the annual time series estimates for the average value of land in German counties and county-free cities. As described in the previous

Section, for each county, we use a two step estimation process following Davis and Palumbo (2008).

For our dataset, we use house price indexes for single- and double-family houses on the county and independent city level for the years 2007-2018 from vdp Research GmbH⁶. For the construction costs at the county level, we use data published by the BKI⁷. First, we use construction cost data per square meter of single- and double-family houses from 2016 (BKI, 2016). Second, we use BKI construction cost regional factors for German counties from 2012 to 2019⁸.

In the first step, we combine microdata for key housing variables from the ImmobilienScout24 website with data on construction cost from the Construction Cost Information Centre of German Chambers of Architects GmbH (BKI)⁹. The goal of this first step, for each county, is to generate a benchmark estimate of the replacement cost of single-/double-family, owner-occupied residential structures and a corresponding share of home values that are represented by the value of residential structures. ¹⁰

For our estimation, we use a total of 42,685 observations that met all of our selection criteria. Table 1 shows the summary statistics, where the median number of observations for each county is 106.5, with a minimum sample of 4 and a maximum sample of 1262. We drop the observations of 23 of 402 counties with less than 10 observations per county. Table 5 in Appendix 6.3 lists the raw number of observations by county.

	Number of observations			
number of counties	mean	sd	\min	max
402	106.5746	106.193	3	1007

Table 1: Number of observations by county - descriptive statistics

To generate our benchmark estimates for replacement costs for each housing unit in the sample, we impute the cost of rebuilding the structure as if it were brand new as of the year the ad was removed from the ImmobilienScout24 website. And hence, we estimate a cost function for new single-/double-family detached homes using construction cost data published by the BKI (2016). We then apply this cost function to each of the housing units in the sample.

We use the following estimated cost-function that captures the essential variations in costs per square meter across different types:

New cost per square meter, German average,
$$2016 = (5)$$

 $2801.298 + 200.3204 * \text{basement} - 0.789 * \text{hgsize} - 497.840 * I(\text{floors} \ge 2)$

-705.377 * I(quality = 1) - 274.561 * I(quality = 2)

 $^{^{6}\}mathrm{The}$ time series from 2007 onwards are based on evaluations of property databases compiled by the vdp Research.

⁷BKI stands for Baukosteninformationszentrum Deutscher Architektenkammern.

⁸After 2011, the method of calculating the regional construction cost factors was updated. Hence, according to experts from the surveyor commissions, it would not be consistent to use a longer time series.

 $^{^9\}mathrm{Baukosteninformationszentrum}$ Deutscher Architektenkammern GmbH

 $^{^{10}\}mathrm{More}$ detailed description of methods and variables that we use is in Appendix 6.1

3 METHODS AND DATA

In Equation (5), $I(\cdot)$ is an indicator function that is 1 if the expression is true and 0 otherwise¹¹.

Using a depreciation rate for the new building cost for structure, the following $s_{i,t}$ denotes the replacement cost of the structure after accounting for depreciation:

$$s_{i,t} = str_cost_new_{i,t} \cdot \left(\frac{1}{1+\delta}\right)^{age_{i,t}}$$
(6)

where, $str_cost_new_{i,t}$ denote the new building cost for structure of housing unit *i* in period *t* (using Equation 5). The variable $age_{i,t}$ is the age (in years) of the structure of housing unit *i* in period *t*. δ is the annual depreciation rate, which we set equal to 0.02.¹².

Lastly, we calculate a county-wide benchmark structures share for the year 2012 as it is the first year of our time series and denote it as w_t^s in period t = 2012:

$$w_{2012}^s = \sum_i \frac{s_{i,2012}}{value_{i,2012}},\tag{7}$$

where we calculate $s_{i,2012}$ by

$$s_{i,2012} = s_{i,eyear} * \frac{RF_{2012,j} * CCI_{2012} * CPI_{2012}}{RF_{eyear,j} * CCI_{eyear} * CPI_{eyear}}$$
(8)

and $value_{i,2012}$ by

$$value_{i,2012} = value_{i,eyear} * \frac{HPI_{2012,j} * CPI_{2012}}{HPI_{eyear,j} * CPI_{eyear}}.$$
(9)

The index *i* indicates a housing unit and *j* refers to a specific county. RF denotes the regional construction cost factors and CCI the Germany-wide construction cost index. CPI refers to the consumer price index and HPI is the house price index on county-level from vdpResearch. The summation in both the numerator and denominator in Equation (7) is over all housing units/housholds in a particular county. According to the assumption that a home's value is the sum of the replacement cost of the physical structure and the market value of the land, it follows a benchmark land share for a given county of $w_t^l = 1 - w_t^s$.

3.2 Extrapolating the Benchmark Estimates for Structures Share and Replacement Cost: German Case

To estimate a continuous annual time-series of structure shares, we extrapolate forward the benchmark structures share derived above: This is the second step of the two-step procedure. Recall that the total value of housing in a particular county at date t is defined as:

$$p_t^h h_t = p_t^l l_t + p_t^s s_t, (10)$$

 $^{^{11}\}mathrm{More}$ details on the cost function estimation are in Appendix 6.1.

 $^{^{12}}$ We follow the linear depreciation schedule for tax return, where the starting point for calculating depreciation on buildings is the acquisition or construction cost of the building. The proportionate land costs are not included in the basis of assessment. See, https://www.steuertipps.de/lexikon/a/abschreibung-gebaeude

3 METHODS AND DATA

where the market value of land is denoted as $p_t^l l_t$ and the overall replacement cost in that county in a given period as $p_s^s s_t$. Furthermore, Davis and Palumbo (2008) define the total nominal value of structures in a particular county at period t + 1 as the sum of the total nominal value in the previous period t, revalued for changes in construction costs $(\frac{p_{t+1}^s}{p_t^s})$ and the nominal net new structures $(p_{t+1}^s \Delta s_{t+1})$, i.e. new structures less depreciation:

$$p_{t+1}^{s}s_{t+1} = p_{t}^{s}s_{t}\left(\frac{p_{t+1}^{s}}{p_{t}^{s}}\right) + p_{t+1}^{s}\Delta s_{t+1}$$
(11)

Moreover, we assume that the nominal value of net new structures in a particular county is equal to some proportion (θ_t) of the nominal value of net new housing in a particular county $(p_{t+1}^h \Delta h_{t+1})$:

$$p_{t+1}^s \Delta s_{t+1} = \theta_t p_{t+1}^h \Delta h_{t+1} \tag{12}$$

Inserting (12) in (11) and dividing by the nominal value of housing at t + 1, $p_{t+1}^h h_{t+1}$, we have the share of structure as

$$\frac{p_{t+1}^s s_{t+1}}{p_{t+1}^h h_{t+1}} = \frac{p_t^s s_t}{p_t^h h_t} \frac{\frac{p_{t+1}^s}{p_t^s}}{\frac{p_{t+1}^h}{p_t^h}} \left(\frac{h_t}{h_{t+1}}\right) + \theta_t \frac{\Delta h_{t+1}}{h_{t+1}},\tag{13}$$

where the identity $p_{t+1}^h h_{t+1} = p_t^h h_t \frac{p_{t+1}^h}{p_t^h} \frac{h_{t+1}}{h_t}$ is used. When we let the total structures share of aggregate house value in a county in period t be defined as $w_t^s = \frac{p_t^s s_t}{p_t^h h_t}$ and inserting into equation (3), we obtain Equation (4) or

$$w_{t+1}^{s} = w_{t}^{s} \left(\frac{\frac{p_{t+1}^{s}}{p_{t}^{s}}}{\frac{p_{t+1}^{h}}{p_{t}^{h}}} \right) \left(\frac{h_{t}}{h_{t+1}} \right) + \theta_{t} \frac{\Delta h_{t+1}}{h_{t+1}}.$$
 (14)

Given structures share in period t, w_t^s , the growth rates in construction cost and house prices, $\frac{p_{t+1}^s}{p_t^s}$ and $\frac{p_{t+1}^h}{p_t^h}$, a value for the structure intensity of nominal net new housing, θ_t , and a proxy for the growth rate of real housing stock, $\frac{h_t}{h_{t+1}}$, we can now calculate the structures share in period t+1, w_{t+1}^s using Equation (14).

To implement this dynamic equation we use the benchmark structures share for each county in 2012 derived in the previous section. For growth in construction cost by county, we multiply county specific time-series of regional construction cost factors from the BKI with the Germany wide construction cost index for new residential buildings from the German Federal Statistical Office. Both indices are annual time-series.¹³ For growth in house prices, we use constantquality county-specific house prices for single-/double family houses from vdpResearch, based on property transaction database carried out by vdpResearch.

¹³We use the "Baupreisindex" series of the German Federal Statistical office, which is a constant qualityindex and should include builder's margins. We could alternatively use the "Baukostenindex" which excludes changes in productivity and builder's margins. Doing so, those changes however would be attributed to the value of land instead. https://www.destatis.de/DE/Themen/Wirtschaft/Preise/Baupreise-Immobilienpreisindex/Publikationen/Downloads-Bau-und-Immobilienpreisindex/bauwirtschaft-preise-2170400193224.pdf?__blob=publicationFile

4 RESULTS

We approximate the growth in the real housing stock $\left(\frac{h_{t+1}}{h_t}\right)$ by the growth in the number of households of a county. We assume that both growth rates are proportional. The number of households by county in each period is computed by dividing the population in a particular county by the average household size of Germany in that period¹⁴. We use population data by county and the number of households of a certain size in Germany from the federal statistical office, where the latter is used to construct the average household size in Germany in a given period. Both datasets are reported at an annual frequency.

Lastly, we assume that the fraction of new home value accounted for by new structures is

$$\theta_t = \frac{exp(1.98w_t^s)}{1 + exp(1.98w_t^s)}.$$
(15)

The parameter, θ , allows for varying land intensity of new housing with the average land intensity in a county. We choose the scale parameter to be 1.98 to have a structure's share of new housing of around 0.77 when the average structures share in a county is 0.6.

We do not have an estimate for the share of the nominal value of structures in percent of the nominal value of new housing as Davis and Palumbo (2008). However, we assume a value of around 80%, arguing that land prices increased more than construction costs. Suggesting an increased land intensity of new housing (Figure 2).¹⁵ Moreover, we still choose a relatively high value as construction style is quite more expensive in Germany compared to the US. ¹⁶

4 Results

Figure 3 shows the total number of observations on home values that we construct for 379 of 402 German counties and county-free cities using the annual dataset from 2014 and 2017.¹⁷

¹⁴There is no continuous data on average household size by county. Although we are aware that household size varies across counties, we use this assumption as the estimation method only requires percentage changes in the average household size by the county to be correct.

¹⁵Note that the housing price index was slightly decreasing from 2000 to 2008, although both the land price index and construction costs have been increasing. We think this counter-intuitive fact is partially due to that the land prices represent only for the "ready to build" land. Consequently, the prices for the non-useable construction land could have been decreasing in those periods: These land prices are not publicly available.

 $^{^{16} \}rm https://www.nwb-experten-blog.de/kaufpreisaufteilung-fuer-ein-bebautes-grundstueck-darf-es-ein-bisschenmehr-gebaeude-sein/$

¹⁷We drop counties where we have less than 10 observations from the ImmobilienScout24 dataset (Indicated by a yellow outline in Figure 2).

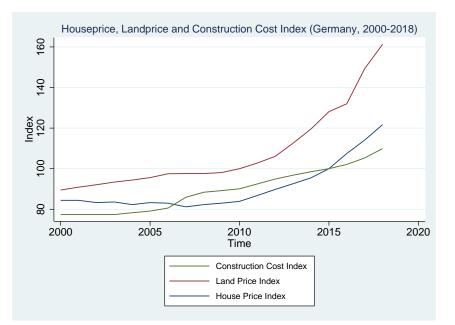


Figure 2: German Housing Market Development; Source: Destatis, German Federal Statistical Office



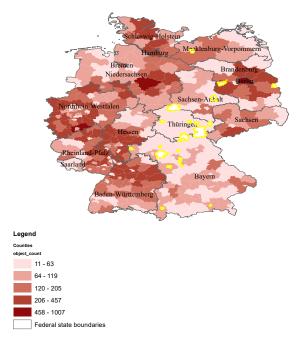


Figure 3: Number of observations by county

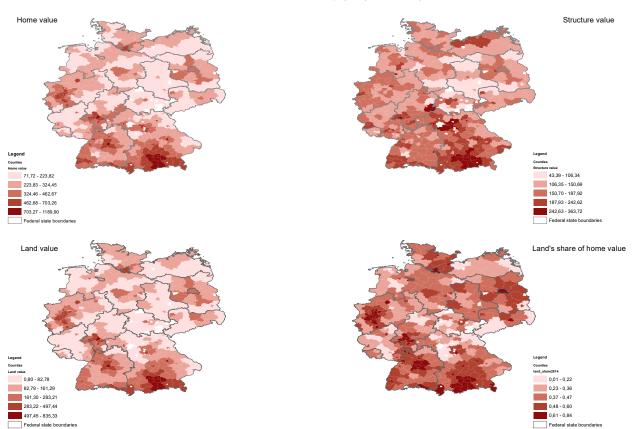
Figure 4 shows the components of home value by county in 2014. We construct price indices for residential land and estimate average values for the stock of single-/double-family owneroccupied housing. We also estimate their structure and land components by estimating land's share of home value. We present a detailed description in Figure 4 below by the state level, differentiating between the new federal states¹⁸ and the old federal states of Germany. We then describe the basic differences among the German states and counties and highlight some of the trends over time.

4.1 Components of Home Value in 2014

Table 2 shows the home, land and structure values in $1000s \in$ as well as the land's share of home value and the land value per square meter for German federal states, divided into East and West. On average, homes are more expensive in West German states compared to East German states in 2014. The city-states of Hamburg and Berlin, followed by the southern German states of Bavaria and Baden-Württemberg, have the by far highest home values in 2014. This pattern also becomes evident in Figure 4. Home values and land values are significantly higher in the south of Germany (especially in Bavaria around Munich and in Baden-Württemberg around Stuttgart). But also the large agglomeration of urban centers in the Ruhr area (Nordrhein-Westfalen) and Hamburg show high home and land values.

Although there are quite significant differences in the value of residential land, the states and corresponding counties differ relatively little in terms of the average replacement cost of structures. Two features that highlight the difference between the West and East German states are first that the land's share is significantly higher in the West than East (51% versus 38%). Second, Table 2 also shows that the land value is significantly higher than the structure in the West (184 versus 158), whereas, in the East, the structure value is higher than the value of land (82 versus 132). At the end of 2014, residential land accounted for between 27 and 48 percent in states of former East Germany and between 36 and 70 percent of home value in Western German states, where Hamburg has the highest share of land in home value, with 70 percent.

 $^{^{18}\}mathrm{Also}$ known as "East Germany" and before 1990 part of the German Democratic Republic(DDR).



Components of home value by geographic region 2014

Figure 4: Components of home value by county in 2014

Region/State	Home value (1000s Euros)	Land value (1000s Euros)	Structure value (1000s Euros)	Land's share of home value (pct)	Land value per square meter
West	342	184	158	51	253
Schleswig-Holstein	278	135	144	48	159
Hamburg	600	420	180	70	546
Niedersachsen	227	87	140	38	103
Bremen	242	101	141	42	135
Nordrhein-Westfalen	335	181	154	54	242
Hessen	357	183	174	51	286
Rheinland-Pfalz	262	112	150	43	177
Baden-Württemberg	406	231	175	57	349
Bayern	425	223	203	52	307
Saarland	200	73	127	36	95
Berlin	434	280	154	64	387
East	214	82	132	38	92
Brandenburg	248	118	130	48	131
Mecklenburg-Vorpommern	239	79	160	33	86
Sachsen	239	89	151	37	98
Sachsen-Anhalt	175	77	98	44	87
Thüringen	167	45	122	27	57

Table 2: Com	ponents of	home	value	in	2014	by	state
--------------	------------	------	-------	----	------	----	-------

4.2 Changes in Home Value: 2014 - 2017

In Table 3 and Figure 5, we present cumulative changes in the components of home value and land's share of home value between 2014-2017 in real terms, i.e., converted to 2015 Euros using the Consumer Price Index from the German federal statistical office, by state and county, respectively. The changes, on average, in the land value for both West and East Germany are much more significant than the changes in the home values: 30% versus 108% in the West, and 15% versus 91% for the East. On the other hand, the structure value increased only marginally, 7% and 12%, for the West and East, respectively. The cummulative change in land values varies between 9% and 171% (excluding Berlin with exorbitant high increases of 668% ¹⁹ and Saarland with a decrease of 18%). However, replacement costs only vary between -5% and 26% (again excluding Berlin with an increase of 41%).

¹⁹We emphasize that our goal is to estimate land prices, and hence, we are silent about the reasons for the land price changes in our work. But such an increase in Berlin's land values needs a bit of attention. A recent work by Möbert (2020) indicates that, among other factors, Berlin's population growth contributes to a significant increase in demand for housing and land. Berlin has experienced both inward migration (at an average of 40 000 new inhabitants annually) and non-Germans migration (at an average increase of 8.3% foreign nationals in Berlin annually) since 2011. One other factor that could contribute to Berlin's extraordinary land value increase in the last few years is the inflow of foreign investments. According to Schaer (2018), up to 68 percent of Berlin apartments were sold to foreigners in 2015.

	Cumulative change in					
Region/State	Home value (pct)	Land value (pct)	Structure value (pct)	Land's share of home value (pctg pts)	Land value per square meter	
West	30%	108%	7%	9	125%	
Schleswig-Holstein	9%	29%	4%	4	28%	
Hamburg	46%	82%	-4%	3	93%	
Niedersachsen	6%	26%	5%	17	29%	
Bremen	20%	80%	18%	8	66%	
Nordrhein-Westfalen	7%	27%	2%	8	28%	
Hessen	11%	37%	5%	8	48%	
Rheinland-Pfalz	9%	140%	5%	29	163%	
Baden-Württemberg	16%	62%	5%	7	86%	
Bayern	15%	55%	5%	5	65%	
Saarland	-12%	-18%	-5%	4	-20%	
Berlin	207%	668%	41%	9	790%	
East	15%	91%	12%	1	104%	
Brandenburg	21%	43%	17%	4	65%	
Mecklenburg-Vorpommern	16%	28%	12%	1	39%	
Sachsen	10%	202%	2%	3	222%	
Sachsen-Anhalt	2%	9%	2%	0	10%	
Thüringen	24%	171%	26%	-5	186%	

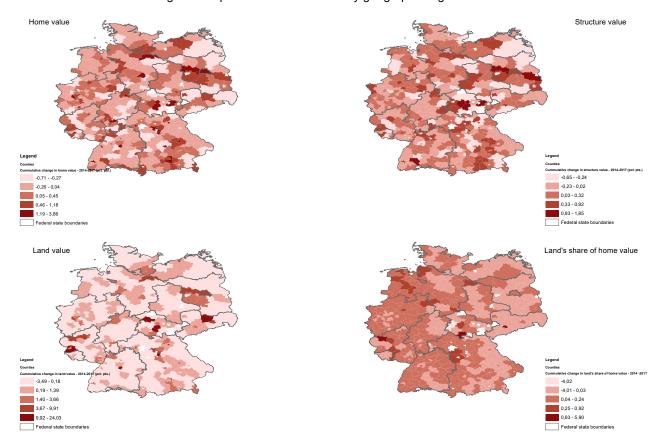
Table 3: Change in components of home value by state - 2014 through 2017

While values in Table 2 tend to vary quite significantly between East and West German states, the cumulative changes in Table 3 appear to be evenly distributed accross both Eastern and Western German states. However, land's share of home value increases more in the Western parts of Germany compared to the new federal states (East). Moreover, one of the most notable features in Figure 5^{20} is that the home values for urban centers such as Hamburg, Berlin, and Munich increased quite substantially more than the surrounding counties. The result reinforces the monocentric Bid-Rent theory that the home and land prices (rents) are inversely related to the distance away from the central business district (CBD).

Moreover, the results in Table 3 indicate that the land component is the main driver for the housing price increase between 2014 and 2017 in Germany.²¹ Most regions experienced a significant increase in home value and the share of the market value of residential land in home value. At the same time, the change in the replacement cost of structures was comparatively small.

 $^{^{20}\}mathrm{Note:}$ Classifications of the breaks vary from map to map in Figure 5.

 $^{^{21}\}mathrm{Our}$ estimation results for the components of home value for 2017 is in Appendix 6.2.



Change in components of home value by geographic region - 2014-2017

Figure 5: Change in components of home value by county - 2014-2017

4 RESULTS

4.3 Changes in the Distribution of Land's Share of Home Value and Residential Land Values; 2014-2017

Figure 6 presents the cumulative distribution function of land's share of home value across the 379 German counties and county-free cities in our sample as of 2014 and 2017. Figure 6 shows a relatively even rightward shift of around 4 percent for the second to the ninth percentile of the distribution. The lowest and highest 5 percent of the distribution changed if at all only very slightly. This change suggests that the counties at the top and bottom only shuffled their order, whereas counties in the middle of the distribution experienced an overall rightward shift in the land's share of home value. The range from lowest (around 7 percent) to highest (around 85 percent) land's share of home value stays quite the same between the years. Consistent patterns are shown for the cumulative distribution function of the value of residential land, which is displayed in Figure 7 across the 379 German counties in our sample as of 2014 and 2017.

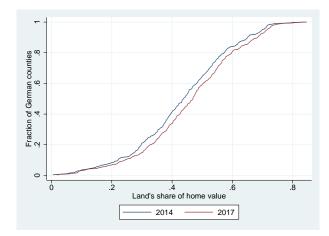


Figure 6: Cumulative distribution of land's share of home value across our sample of 379 German counties

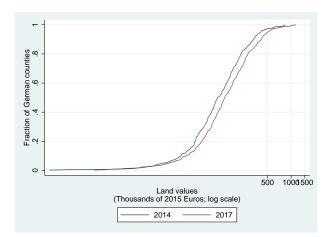


Figure 7: Cumulative distribution of land values across our sample of 379 German counties

4.4 Changes in the Distribution of Land value and Comparison of our Land Value Estimates

Figure 8 maps land values per square meter of lot size for German counties in 2014 and 2017.²² As expected, the estimated land values per square meter of lot size are highest in counties in and around large agglomerations and decrease the further away from those one gets. Figure 9 shows our - and the expert-based - land value estimates for Hessen's counties in 2014.²³ We choose the state of Hessen for our comparison because the state provides the most available dataset. While expert-based values suggest lower land values in the county-free city of Kassel and higher values in the county of Kassel, our estimates suggest that land values are higher in the city and lower in the county around the city of Kassel. Moreover, our estimates suggest that the land values for Wiesbaden are inversely related to the distance from Frankfurt. In contrast, the expert-based land values are significantly higher also for the county Darmstadt-Dieburg. Figure 9 clearly shows that the differences between our and the experts estimates are not trivial and could have large consequences in the land price valuation.

 $^{^{22}\}mathrm{The}$ cut of levels between classes for the year 2017 are adjusted to those of 2014.

 $^{^{23}}$ We adjust the cut off levels between classes for the expert-based values to those of our estimates.

Land value/ square meter

2014

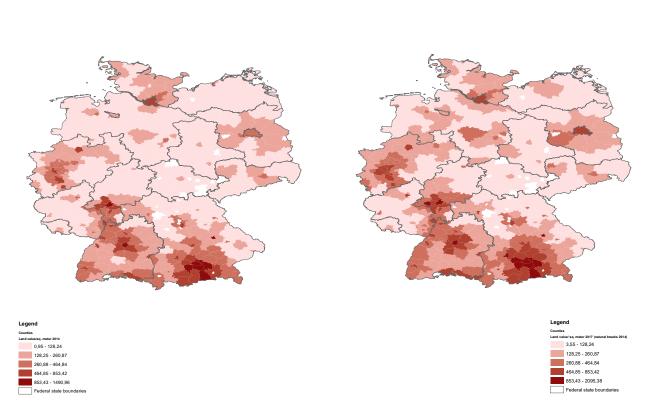
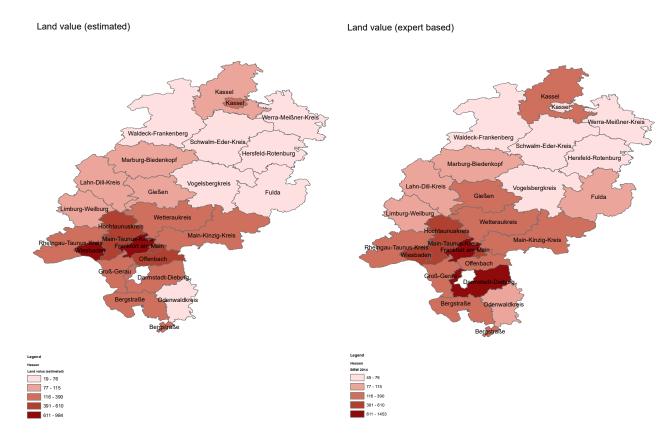


Figure 8: Estimated Land values per square meter of lotsize by county in 2014 and 2017

2017



Expert based land values vs estimated land values per square meter of lot size - Hessen 2014

Figure 9: Estimated Land values vs Expert based land values per square meter of lotsize in Hessen 2014

5 Discussion/Conclusion

We estimate land prices and construct a new database for the values of residential land across 379 of 402 German counties and county-free cities over time. Using a total of 42,685 observations from 2014 to 2017, we show that the price of residential land has become relatively more expensive in the majority of German counties. As the urban theory predicts, the highest home and land price increases are in counties around urban centers Munich, Stuttgart, Berlin, Hamburg, Dresden, and the cities in the Ruhr area. But, in general, almost all the counties show an upward shift in home -, land values, the share of residential land in home values. The most significant differences are and seem to remain between the new and old federal states of Germany. The Eastern states are in par with the West (excluding Berlin) in terms of cumulative percentage changes in the value of the components of the housing bundle. However, there was a significant difference in the average land's share of home value increased between East and West. Moreover, the absolute differences in home values, the value of residential land, and replacement costs did not decrease between the two parts of Germany.

Our results show that a more significant share of land in home value could lead to faster home price appreciation in the future. Consequently, one of the implications of our results is that real land prices drive the movements in home prices and the cycles in the value of land are likely to influence the evolution of house prices more in the future than they did in the past. This implication would affect urban areas with a larger inelasticity of supply in housing. Moreover, this result could lead to more significant swings in house prices due to changes in demand.

In our work, we are silent on some of the shortcomings of the residual method. First, we could use the simple option model as in Davis et al. (2019) to address the assumption that the value of a home is equal to the sum of the replacement cost of its structure and the value of the land. Excluding houses that have an "effective age" of more than 20 years. Second, we could use a more extensive transaction data set to entirely avoid depreciation of replacement costs by using only new homes to estimate the share of land in home value (e.g., Davis et al., 2017).

6 Appendix

6.1 Details on the Benchmark Estimates for the Replacement Cost of Residential structures: German Case

In this section, we provide further detailed description of our approach in computing the annual time series estimates for the average value of land in German counties and county-free cities.

In the first step, we use the house price listings from the internet platform Immobilien- $Scout24^{24}$ with the RWI - Leibniz Institute for Economic Research as well as the 2016 average construction cost data per square meter of single- and double-family houses. We use the following set of variables from the dataset that characterizes "house": *built* (the year of construction),

 $^{^{24}}$ See Boelmann and Schaffner (2019) for data documentation.

basement (whether the housing unit has a basement), garage (the availability of parking lots)²⁵, hgsize (the living size of the dwelling in square meters), lotsize (the lot size corresponding to the dwelling in square meters), floors (counts the number of floors), rooms (counts the number of rooms), value (denotes the posted selling price of the advertisement of the housing unit), and quality (records the quality of a building's facilities). We categorize the reported qualities into "simple", "normal" and "high", where the latter is combined of the two highest categories in the variable facilities.

We use all the listed single-/double-family houses for sale using the variables "house_category" and "real estate type". We, however, exclude semi-attached single-family buildings ("Doppel-haushälften") from our dataset as we cannot identify whether construction costs refer the one or both parts of the duplex ("Doppelhaus").²⁶ We also exclude all the observations that are listed as "unfinished buildings", i.e., we exclude objects that are still under construction but are listed and the observations that refer to as "holiday homes".

Although energy-efficient construction is prevalent and subsidized by the German federal government,²⁷ contrary to our expectations we find energy-efficiency standards not to have a significant impact when estimating our construction cost function. Hence, we do not include a variable *energy* that depicts the energy-efficiency standards of the houses. To eliminate duplicates, we use the variable $dupid_gen$ generated by the RWI that classifies object identifiers used more than once. We remove those observations with a duplicate ID of 1, indicating that they are identical in essential variables and were posted several times around the same date. Also, we drop objects with duplicate identifier 2, i.e., close to identical objects as above; however, we keep objects that differ in essential variables and those with object identifiers used only once.

To eliminate possible scam postings and very luxury objects, we eliminate extreme outliers and unrealistic values in the variables we use. First, we remove houses with more than 600 or less than 50 square meters living size, more than 15 rooms and more than 2500 square meters of lot size that are possible for agricultural use. Second, we drop observations with values of less than 10,000 or more than 5,000,000 Euros.²⁸

For our cost function estimation, we use the BKI (2016) that publishes how much it would cost per square meter of living space to build a new single-/double-family home on average in Germany in 2016, given several characteristics of the structure. The characteristics published by the BKI include total square meter of living space, the number of stories²⁹, the veneer/siding, the quality of the building's facilities, the energy-efficiency standard, and whether it has a basement or not. We draw our analysis upon square-foot building costs of single-/double-family homes in the BKI data. Using the sample houses from the BKI database, we have 106 observations for

²⁵Unfortunately, no variable explicitly states the kind of parking lot. However, we think it is reasonable to assume that a substantial fraction of single-/double-family dwellings in Germany is equipped with a garage.

 $^{^{26}}$ Effenberger (2015) reports that semi-attached single-family buildings ("Doppelhaushälften") make up around 33.5% of the stock of single family homes in Germany in 2012. But given the construction cost issue, we exclude these stocks.

²⁷See https://www.bmi.bund.de and https://www.kfw.de

 $^{^{28}\}mathrm{We}$ follow the restrictions set by the Immobiliens cout24-House-Price-Index.

 $^{^{29}}$ As there is a picture attached to each housing unit, we can infer these information.

estimating our cost function.³⁰

To obtain total construction cost for each housing unit reported in our sample, we multiply the construction cost per square meter using our cost function by the reported unit size in square meters and add 10,000 for units having a garage.³¹ As this current measure of total construction cost is the German average for the year $2016q1^{32}$, we multiply the cost by the following ratio to normalize the construction cost to reflect for the "end of advertisement"-year and a given county.

BKI Regional factor of the county * Construction Cost Index, "end of advertisement"-year German average of BKI Regional factors* Construction Cost Index, 2016 (16)

After the cost adjustment by Equation (16), we depreciate the structure cost based on its age to better estimate the actual replacement cost of the structure, i.e., if the land had no value. Following Davis and Palumbo (2008), one may think about depreciation as the expense that it would cost to renovate the existing structure to "like-new" standards. Consequently, our replacement cost of the structure after accounting for depreciation is

$$s_{i,t} = str_cost_new_{i,t} \cdot \left(\frac{1}{1+\delta}\right)^{age_{i,t}}$$

6.2 Components of Home Value in 2017

Table 4 and Figure 10 display values of the components of the housing bundle in 1,000s \in as well as the land's share of home value in percent and the land value per square meter of lot size for German federal states and counties, respectively. Comparing Figure 4 and Figure 10 we see that differences between counties stayed almost the same. However, Figure 11 ³³ and Table 4 indicate that especially home and land values became more expensive compared to 2014. Especially the counties around the prosperous urban centers in the Southern states of Germany and Nordrhein-Westfalen became more expensive in terms of home values and land values, but also values around the city states Berlin and Hamburg (Figure 10). The most affordable home values and land values remain to be in the central parts of Germany and in some regions in the East of Germany (Figure 3)) and 732,000 \in in Hamburg. With average home values of around 500,000 \in Berlin, Bavaria and Baden-Württemberg follow the leader in terms of expensive home values. The same picture results for land's share of home values. The average lot was worth between 26 percent of a home's value in Thüringen and 71 and 73 percent in Berlin and

 $^{^{30}}$ The variables are defined as varying in home size (from about 90 - 350 square meters of living space), quality (low, normal/average, high), veneer/siding (wood, brick, man-made siding), basement (yes and no), and two-height dimensions (more or less than two stories).

³¹We base our number 10,000€ from the website that reports the standard sizes for single-/double-garages (https://www.garagen-welt.de/blog/garagengroesse-garagenbreite.html). We then multiply the standard size of a single- and double-garage in m³ with average building costs for a middle-sized single-floor garage in $€/m^3$. See the following document that outlines the garage costs. https://docplayer.org/51789280-Richtwerte-fuer-die-ermittlung-der-verschiedenen-baukosten-als-grundlage-der-festsetzung-von-baugenehmigungsgebuehren.html.

 $^{^{32} \}mathrm{Inclusive}$ of 16% value-added tax.

 $^{^{33}}$ In Figure 11 we adjusted the cut-off values in the legend to the corresponding values of 2014.

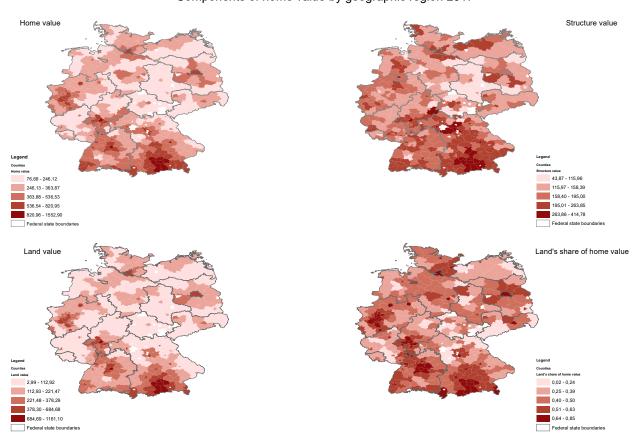
6 APPENDIX

24

Hamburg, respectively. While, the counties in the new federal states of Germany do not lag behind in terms of cummulative changes in home value and the value of residential land, the cross regional differences remain.

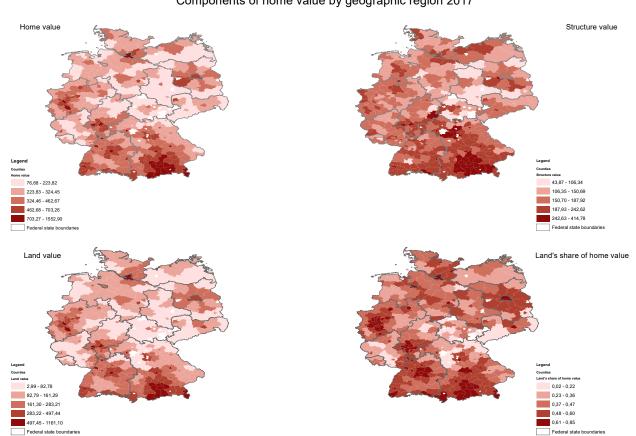
Region/State	Home value (1000s Euros)	Land value (1000s Euros)	Structure value (1000s Euros)	Land's share of home value (pct)	Land value per square meter
West	409	235	173	54	324
Schleswig-Holstein	321	159	161	50	189
Hamburg	732	533	199	73	692
Niedersachsen	262	113	149	43	133
Bremen	276	125	151	45	168
Nordrhein-Westfalen	387	222	164	57	296
Hessen	419	232	187	55	363
Rheinland-Pfalz	299	139	160	46	220
Baden-Württemberg	477	284	193	60	431
Bayern	503	276	227	55	381
Saarland	220	79	141	36	103
Berlin	601	426	175	71	588
East	235	92	143	38	104
Brandenburg	281	138	143	49	154
Mecklenburg-Vorpommern	266	94	173	35	103
Sachsen	262	97	165	37	108
Sachsen-Anhalt	188	83	105	44	95
Thüringen	177	47	130	26	60

Table 4: Components of home value in 2017 by state



Components of home value by geographic region 2017

Figure 10: Components of home value by county in 2017



Components of home value by geographic region 2017

Figure 11: Components of home value by county in 2017 - legend adjusted to 2014

6 APPENDIX

6.3 Table

Region	State	County	Number of observations
West	Schleswig-Holstein	SK Flensburg	37
		SK Kiel	49
		SK Lübeck	47
		SK Neumünster	34
		LK Dithmarschen	152
		LK Herzogtum Lauenburg	121
		LK Nordfriesland	86
		LK Ostholstein	116
		LK Pinneberg	214
		LK Plön	100
		LK Rendsburg-Eckernförde	358
		LK Schleswig-Flensburg	257
		LK Segeberg	217
		LK Steinburg	107
		LK Stormarn	128
	Hamburg	SK Hamburg	223
	Niedersachsen	SK Braunschweig	99
		SK Salzgitter	111
		SK Wolfsburg	158
		LK Gifhorn	457
		LK Göttingen	85
		LK Goslar	109
		LK Helmstedt	178
		LK Northeim	88
		LK Osterode am Harz	101
		LK Peine	163
		LK Wolfenbüttel	183
		LK Region Hannover	917
		LK Diepholz	72
		LK Hameln-Pyrmont	204
		LK Hildesheim	214
		LK Holzminden	65
		LK Nienburg (Weser)	64
		LK Schaumburg	172
		LK Celle	214
		LK Cuxhaven	118
		LK Harburg	161
		LK Lüchow-Dannenberg	126
		LK Lüneburg	134
		LK Osterholz	76
		LK Rotenburg (Wümme)	104
		LK Soltau-Fallingbostel	121
		LK Stade	109
		LK Uelzen	116

Table 5: Observations used to calculate benchmark structures share

	LK Verden	63
	SK Delmenhorst	45
	SK Emden	14
	SK Oldenburg	81
	SK Osnabrück	68
	SK Wilhelmshaven	17
	LK Ammerland	74
	LK Aurich	94
	LK Cloppenburg	63
	LK Emsland	87
	LK Friesland	50
	LK Grafschaft Bentheim	51
	LK Leer	64
	LK Oldenburg	59
	LK Osnabrück	153
	LK Vechta	27
	LK Wesermarsch	26
	LK Wittmund	46
Bremen	SK Bremen	141
	SK Bremerhaven	18
Nordrhein-Westfalen	SK Düsseldorf	91
	SK Duisburg	98
	SK Essen	85
	SK Krefeld	71
	SK Mönchengladbach	104
	SK Mülheim an der Ruhr	59
	SK Oberhausen	41
	SK Remscheid	55
	SK Solingen	60
	SK Wuppertal	115
	LK Kleve	256
	LK Mettmann	186
	LK Rhein-Kreis Neuss	314
	LK Viersen	280
	LK Wesel	$\frac{200}{334}$
	SK Bonn	88
	SK Köln	163
	SK Leverkusen	50
	SK Städteregion Aachen	227
	LK Düren	191
	LK Rhein-Erft-Kreis	256
	LK Euskirchen	$230 \\ 247$
	LK Heinsberg	259
	LK Oberbergischer Kreis	239 296
	LK Rheinisch-Bergischer	$\frac{230}{236}$
	Kreis	200
		581
	LK Rhein-Sieg-Kreis SK Bettrop	$\frac{581}{29}$
	SK Bottrop SK Gelsenkirchen	
		20 50
	SK Münster	59

Hessen

LK	Borken	205
LK	Coesfeld	155
LK	Recklinghausen	233
$\mathbf{L}\mathbf{K}$	Steinfurt	229
LK	Warendorf	101
SK	Bielefeld	113
LK	Gütersloh	78
LK	Herford	152
LK	Höxter	103
LK	Lippe	385
LK	Minden-Lübbecke	253
LK	Paderborn	136
SK	Bochum	32
SK	Dortmund	124
SK	Hagen	55
	Hamm	31
SK	Herne	11
LK	Ennepe-Ruhr-Kreis	153
	Hochsauerlandkreis	123
LK	Märkischer Kreis	235
LK	Olpe	54
	Siegen-Wittgenstein	203
	Soest	139
LK	Unna	224
SK	Darmstadt	33
SK	Frankfurt am Main	112
SK	Offenbach am Main	24
SK	Wiesbaden	97
LK	Bergstraße	248
	Darmstadt-Dieburg	231
	Groß-Gerau	189
LK	Hochtaunuskreis	357
LK	Main-Kinzig-Kreis	366
	Main-Taunus-Kreis	165
LK	Odenwaldkreis	164
LK	Offenbach	228
	Rheingau-Taunus-Kreis	252
	Wetteraukreis	391
LK	Gießen	215
LK	Lahn-Dill-Kreis	274
	Limburg-Weilburg	292
	Marburg-Biedenkopf	100
	Vogelsbergkreis	132
	Kassel	36
	Fulda	57
	Hersfeld-Rotenburg	108
	Kassel	104
	Schwalm-Eder-Kreis	49
	Waldeck-Frankenberg	49
	Werra-Meißner-Kreis	80

Rheinland-Pfalz	SK Koblenz	42
	LK Ahrweiler	140
	LK Altenkirchen	145
	LK Bad Kreuznach	294
	LK Birkenfeld	54
	LK Cochem-Zell	37
	LK Mayen-Koblenz	275
	LK Neuwied	229
	LK Rhein-Hunsrück-Kreis	160
	LK Rhein-Lahn-Kreis	152
	LK Westerwaldkreis	335
	SK Trier	24
	LK Bernkastel-Wittlich	67
	LK Eifelkreis Bitburg-Prüm	83
	LK Vulkaneifel	40
	LK Trier-Saarburg	132
	SK Frankenthal (Pfalz)	22
	SK Kaiserslautern	78
	SK Landau in der Pfalz	31
	SK Ludwigshafen am Rhein	48
	SK Mainz	60
	SK Neustadt an der	41
	Weinstraße	
	SK Pirmasens	36
	SK Speyer	25
	SK Worms	45
	SK Zweibrücken	24
	LK Alzey-Worms	213
	LK Bad Dürkheim	228
	LK Donnersbergkreis	185
	LK Germersheim	124
	LK Kaiserslautern	232
	LK Kusel	88
	LK Südliche Weinstraße	168
	LK Rhein-Pfalz-Kreis	153
	LK Mainz-Bingen	289
	LK Südwestpfalz	141
Baden-Württemberg	SK Stuttgart	89
	LK Böblingen	178
	LK Esslingen	220
	LK Göppingen	158
	LK Ludwigsburg	212
	LK Rems-Murr-Kreis	252
	SK Heilbronn	50
	LK Heilbronn	302
	LK Hohenlohekreis	123
	LK Schwäbisch Hall	219
	LK Main-Tauber-Kreis	54^{210}
	LK Heidenheim	141
	LK Ostalbkreis	285

SK	Baden-Baden	44
SK	Karlsruhe	36
LK	Karlsruhe	266
LK	Rastatt	137
SK	Heidelberg	23
	Mannheim	50
LK	Neckar-Odenwald-Kreis	130
LK	Rhein-Neckar-Kreis	299
SK	Pforzheim	70
LK	Calw	158
LK	Enzkreis	253
LK	Freudenstadt	95
SK	Freiburg im Breisgau	13
LK		83
Bre	isgau-Hochschwarzwald	
	Emmendingen	38
	Ortenaukreis	135
LK	Rottweil	130
LK	Schwarzwald-Baar-Kreis	61
LK	Tuttlingen	91
	Konstanz	73
LK	Lörrach	68
LK	Waldshut	126
	Reutlingen	177
	Tübingen	112
	Zollernalbkreis	139
	Ulm	28
	Alb-Donau-Kreis	148
LK	Biberach	137
	Bodenseekreis	59
LK	Ravensburg	87
	Sigmaringen	69
	Ingolstadt	100
	München	81
	Rosenheim	4
	Altötting	60
	Berchtesgadener Land	20
	Bad Tölz-Wolfratshausen	$\overline{21}$
	Dachau	51
	Ebersberg	52
	Eichstätt	119
	Erding	37
	Freising	53
	Fürstenfeldbruck	79
	Garmisch-Partenkirchen	32
	Landsberg am Lech	52 64
	Miesbach	36
	Mühldorf a.Inn	50 52
	München	106
	Neuburg-Schrobenhausen	84
ъц	reading-bein obeimausen	04

LK	Pfaffenhofen a.d.Ilm	86
LK	Rosenheim	94
LK	Starnberg	70
LK	Traunstein	72
LK	Weilheim-Schongau	47
	Landshut	29
SK	Passau	26
SK	Straubing	13
LK	Deggendorf	147
LK	Freyung-Grafenau	115
LK	Kelheim	69
LK	Landshut	105
LK	Passau	189
LK	Regen	34
LK	Rottal-Inn	55
LK	Straubing-Bogen	59
LK	Dingolfing-Landau	107
	Amberg	17
	Regensburg	30
	Weiden i.d.OPf.	13
LK	Amberg-Sulzbach	60
	Cham	55
LK	Neumarkt i.d.OPf.	57
LK	Neustadt a.d.Waldnaab	31
LK	Regensburg	96
	Schwandorf	131
LK	Tirschenreuth	19
SK	Bamberg	4
	Bayreuth	16
	Coburg	8
	Hof	19
LK	Bamberg	43
	Bayreuth	64
	Coburg	34
	Forchheim	35
LK	Hof	43
LK	Kronach	27
LK	Kulmbach	38
LK	Lichtenfels	11
LK	Wunsiedel	16
i.Fi	chtelgebirge	
SK	Ansbach	3
SK	Erlangen	8
	Fürth	5
SK	Nürnberg	77
	Schwabach	9
	Ansbach	106
LK	Erlangen-Höchstadt	65
	Fürth	41
LK	Nürnberger Land	71

East

	LK	53
	Neustadt/Aisch-B.Windsh.	
	LK Roth	49
	LK	32
	Weißenburg-Gunzenhausen	
	SK Aschaffenburg	9
	SK Schweinfurt	3
	SK Würzburg	11
	LK Aschaffenburg	67
	LK Bad Kissingen	28
	LK Rhön-Grabfeld	15
	LK Haßberge	35
	LK Kitzingen	8
	LK Miltenberg	66
	LK Main-Spessart	27
	LK Schweinfurt	46
	LK Würzburg	69
	SK Augsburg	111
	SK Kaufbeuren	37
	SK Kempten (Allgäu)	8
	SK Mempion (Higau)	$\frac{0}{21}$
	LK Aichach-Friedberg	98
	LK Augsburg	199
	LK Dillingen a.d.Donau	68
	LK Günzburg	62
	LK Neu-Ulm	80
	LK Lindau (Bodensee)	$\frac{30}{28}$
	LK Ostallgäu	55
	LK Unterallgäu	67
	LK Donau-Ries	50
	LK Oberallgäu	83
Saarland	LK Regionalverband	52
Saananu	Saarbrücken	52
	LK Merzig-Wadern	51
	LK Neunkirchen	51 59
	LK Saarlouis	$\frac{39}{39}$
	LK Saarpfalz-Kreis	39 82
	LK St. Wendel	$\frac{32}{36}$
Berlin	SK Berlin	$\frac{30}{1007}$
Brandenburg	SK Brandenburg an der	8
Drandenburg	Havel	0
	SK Cottbus	34
	SK Cottbus SK Frankfurt (Oder)	$\frac{54}{7}$
	SK Potsdam	45
	LK Barnim	$\frac{43}{265}$
		$\frac{205}{276}$
	LK Dahme-Spreewald LK Elbe-Elster	270 60
	LK Havelland	$\frac{60}{185}$
	LK Havenand LK Märkisch-Oderland	
	LK Markisch-Oderland LK Oberhavel	$\begin{array}{c} 250 \\ 240 \end{array}$
		24U

	LK Oberspreewald-Lausitz	64
	LK Oder-Spree	168
	LK Ostprignitz-Ruppin	39
	LK Potsdam-Mittelmark	246
	LK Prignitz	16
	LK Spree-Neiße	51
	LK Teltow-Fläming	125
	LK Uckermark	33
Mecklenburg-Vorpommern	SK Rostock	41
	SK Schwerin	9
	LK Mecklenburgische	115
	Seenplatte	
	LK Rostock	136
	LK Vorpommern-Rügen	49
	LK Nordwestmecklenburg	73
	LK Vorpommern-Greifswald	42
	LK Ludwigslust-Pachim	123
Sachsen	SK Chemnitz	44
	LK Erzgebirgskreis	209
	LK Mittelsachsen	118
	LK Vogtlandkreis	118
	LK Zwickau	127
	SK Dresden	125
	LK Bautzen	106
	LK Görlitz	29
	LK Meißen	144
	LK Sächsische	132
	Schweiz-Osterzgebirge	
	SK Leipzig	111
	LK Leipzig	177
	LK Nordsachsen	83
Sachsen-Anhalt	SK Dessau-Roßlau	4
	SK Halle (Saale)	43
	SK Magdeburg	45
	LK Altmarkkreis Salzwedel	34
	LK Anhalt-Bitterfeld	31
	LK Börde	99
	LK Burgenlandkreis	56
	LK Harz	76
	LK Jerichower Land	30
	LK Mansfeld-Südharz	57
	LK Saalekreis	135
	LK Salzlandkreis	78
	LK Stendal	84
	LK Wittenberg	21
Thüringen	SK Erfurt	22
	SK Gera	8
	SK Jena	7
	SK Suhl	16
	SK Weimar	7

SK	Eisenach	4
LK	Eichsfeld	13
LK	Nordhausen	9
LK	Wartburgkreis	38
LK	Unstrut-Hainich-Kreis	13
LK	Kyffhäuserkreis	13
LK	Schmalkalden-Meiningen	32
LK	Gotha	30
LK	Sömmerda	26
LK	Hildburghausen	24
LK	Ilm-Kreis	28
LK	Weimarer Land	23
LK	Sonneberg	9
LK	Saalfeld-Rudolstadt	22
LK	Saale-Holzland-Kreis	14
LK	Saale-Orla-Kreis	5
LK	Greiz	27
LK	Altenburger Land	11

References

- Albouy, D., Ehrlich, G., & Shin, M. (2018). Metropolitan land values. Review of Economics and Statistics, 100(3), 454–466.
- BKI. (2016). BKI Baukosten 2016 Neubau Teil 1. Statistische Kostenkennwerte für Gebäude. Stuttgart: BKI Baukosteninformationszentrum Deutscher Architektenkammern GmbH.
- Boelmann, B., & Schaffner, S. (2019). Fdz data description: Real-estate data for germany (rwi-geo-red) advertisements on the internet platform immobilienscout24). *RWI Projektberichte*.
- Bostic, R. W., Longhofer, S. D., & Redfearn, C. L. (2007). Land leverage: decomposing home price dynamics. *Real Estate Economics*, 35(2), 183–208.
- Davis, M. A., & Heathcote, J. (2007). The price and quantity of residential land in the United States. Journal of Monetary Economics, 54(8), 2595–2620. doi: 10.1016/j.jmoneco.2007 .06.023
- Davis, M. A., Larson, W. D., Oliner, S. D., & Shui, J. (2019). The price of residential land for counties, zip codes, and census tracts in the united states. *Federal Housing Finance Agency Working Paper*, 19–01.
- Davis, M. A., Oliner, S. D., Pinto, E. J., & Bokka, S. (2017). Residential land values in the washington, dc metro area: New insights from big data. *Regional Science and Urban Economics*, 66, 224–246.
- Davis, M. A., & Palumbo, M. G. (2008). The price of residential land in large US cities. *Journal* of Urban Economics, 63(1), 352–384. doi: 10.1016/j.jue.2007.02.003
- Effenberger, K.-H. (2015). Projektbericht Entwicklungsdynamik EFH Arbeitsbaustein 4: Differenzierung der EFH-Bestände. Leibniz-Institut für ökologische Raumentwicklung.
- Gyourko, J., & Saiz, A. (2006). Construction costs and the supply of housing structure. *Journal* of regional Science, 46(4), 661–680.
- Haughwout, A., Orr, J., & Bedoll, D. (2008). The price of land in the new york metropolitan area. *Current Issues in Economics and Finance*, 14(3).
- Kolbe, J., Schulz, R., Wersing, M., & Werwatz, A. (2012). Location, location, location: Extracting location value from house prices (Tech. Rep.). SFB 649 Discussion Paper.
- Kolbe, J., Schulz, R., Wersing, M., & Werwatz, A. (2015). Identifying berlin's land value map using adaptive weights smoothing. *Computational Statistics*, 30(3), 767-790.
- Kolbe, J., Schulz, R., Wersing, M., & Werwatz, A. (2019). Land value appraisal using statistical methods.
- Möbert, J. (2020). Berlin housing market rent cap may decouple real estate cycle from economic supercycle for a number of years. *Deutsche Bank Research Germany Monitor*.
- Nichols, J. B., Oliner, S. D., & Mulhall, M. R. (2013). Swings in commercial and residential land prices in the united states. *Journal of Urban Economics*, 73(1), 57–76.
- Schaer, C. (2018, August 27). Mayor suggests locking out foreign investors from berlin real estate. *Handelsblatt*.