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Distributional and revenue effects of a tax shift from labor to property

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Distributional and revenue effects of a tax shift from labor to property*

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Abstract

Contrary to frequent recommendations of the public finance literature and international institutions, a persistently high tax wedge on labor is observed in Europe. Simultaneously, the scope for shifting taxes to more growth-friendly revenue sources appears underused. This motivates our simulation of a tax shift from labor to property for Germany, a country where property tax revenues are particularly low and the tax wedge on labor income is among the highest in industrialized countries. We simulate a reform where property is no longer taxed by its (often) outdated cadastral value but by its market value, using the additional revenue to reduce social insurance contributions (SIC). To make such a simulation possible, we match property-related information with the input data of the tax-benefit microsimulation model EUROMOD. We find a considerable increase in property tax revenues, allowing to reduce the implicit tax rate on labor from 37.2% to 36.5%. Distributive effects tend to be modest, and depend critically on the design of the SIC reduction. Overall, our results suggest that more households would gain than lose from the tax shift, with gainers mostly situated in the middle of the income distribution.

JEL: D31, C63, R20, H20

Keywords: labor taxes, property taxes, microsimulations, EUROMOD

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1. Introduction

A high implicit tax rate on labor is often said to be detrimental to growth and employment (e.g. Arnold et al. 2011, Myles 2009). In general, the literature suggests that taxes levied on consumption or property are less distortionary and growth-harming than those levied on corporate or labor income (Mankiw et al. 2009, Slemrod 1990). Despite these findings, the scope for shifting taxes to more growth-friendly revenue sources appears underused in many countries. For instance, various institutions have frequently advised European governments to augment growth potentials by shifting the tax burden away from labor to other tax bases such as property (e.g. European Council 2015, OECD 2014, IMF 2013). Germany in particular has been identified as a country which makes only little use of property taxes,¹ while having a high implicit tax rate on labor (see Figure 1). At the same time, the distribution of income and wealth has become more uneven in many advanced economies (including Germany) over the past few decades, and finding better ways to tax affluent households is back on the policy agenda of many governments (Atkinson & Piketty 2010, Peichl et al. 2010, Bach et al. 2009). Property constitutes the quantitatively most important wealth asset of German households, and the development of real estate prices has been found a crucial component of observed wealth inequalities (Lindner 2015).

The use of outdated cadastral values to determine property tax liabilities is commonly said to be an important reason why revenues from taxing property are so low in Germany (Spahn 2004). Indeed, the current valuation of real estate defining the property tax base dates back to 1964 in Western Germany and to 1935 in Eastern Germany. Various scholars have argued for a revaluation of such cadastral values, but no reform has been carried out (e.g. Blöchliger 2015, Färber et al. 2014). Similar situations with very outdated cadastral values determining property tax liability can be found in several other European countries (Andrews et al. 2011).

Our study first simulates a property tax reform for Germany in which the tax base is no more defined by the cadastral value but by the market value of the property. To assess distributional consequences, we study changes in pre- and post-reform property tax liabilities as well as in disposable income across the income distribution of households. Hence, we follow existing literature recommending the analysis of the joint distribution of income and wealth when interested in distributive effects (e.g. Peichl & Pestel 2013). Further, we simulate two revenue-neutral scenarios in which the additional tax receipts are used to finance a reduction of social insurance contributions on labor income, focusing on first-order effects. Finally, we discuss the different distributional implications of the proposed reform scenarios.

Simulating such a policy reform is difficult since there exists no data source which provides information on both current property tax liability and the actual market value of the property.² However, the HFCS (Household Finance and Consumption Survey)

¹Please note that "property taxes" in this paper describe recurrent levies on immovable real estate owned by private households, i.e. excluding transaction taxes as well as property taxes on corporate assets.

²Only in its survey of 1988, the German Socio-Economic Panel (SOEP) asked respondents about the

of the ECB provides extensive information regarding the value of properties owned. In addition, the EU-SILC survey (European Union Statistics on Income and Living Conditions) contains information on property taxes currently paid. In order to conduct our simulation, we match the two representative survey micro datasets. Performing a number of validity checks we show that especially on a more aggregate level such as household income deciles, the matched dataset preserves the properties of the original HFCS dataset sufficiently well.

The matched dataset is then used to simulate a property tax reform that applies current market values instead of cadastral values. In a first scenario, we assess the potential revenue gain induced by the use of up-to-date property values. Next, we simulate a revenue-neutral scenario in which the additional revenue is used to lower social insurance contributions (SIC) via a lump sum SIC credit. As a third scenario, we simulate a proportional reduction of social insurance contributions, again under revenue-neutrality. All simulations are carried out using EUROMOD – the tax-benefit microsimulation model for EU member states. It allows to evaluate changes in households' disposable income induced by the different scenarios.

From a budgetary perspective, our simulations suggest that the revenue from property taxation would raise from currently $\in 5.8$ bil. to $\in 16.3$ bil. This additional revenue would allow a reduction of the implicit tax rate on labor from currently 37.2 % to 36.5 %. Examining distributive effects, our results first indicate that the (average) percentage increase in the property tax liability is roughly constant across the income distribution of property owners. Hence, the relative size of the property tax liability across the income distribution of homeowners is by and large preserved, which suggests that the reassessment of cadastral values does not hurt a certain income group of property owners in particular. Second, when examining the effect of the proposed update of cadastral values (again *without* redistributing the additional revenue) across the entire income distribution, we find that the relative change in disposable income varies little across income deciles. Thus, an update of cadastral values without using the additional revenue to lower the tax burden on labor would render such a reform virtually neutral in terms of redistribution.

Finally, we turn towards the two revenue-neutral scenarios in which the additional tax receipts are used to lower the tax burden on labor income, considering first-order effects only. We find that when a lump sum SIC credit is granted, all household deciles would gain in disposable income except of the top three ones. In contrast, when using the additional revenue for a proportional reduction of social insurance contributions, the effect on disposable income is small and relatively similar across the income distribution. In sum, we find for both scenarios that more households would gain than lose from the tax shift, with gainers mostly situated in the middle of the income distribution.

Our results relate to existing literature in a number of ways. First, several proposals

cadastral and market values of their main household residence. However, three shortcomings make the use of this joint observation impractical. First, the information dates back to 1988, and property values have changed substantially since then. Second, SOEP only collected ordinal measures of market value. Third and most importantly, information on property is only available for the main household residence and not for any other real estate owned.

have been made to increase tax revenues from wealth and property (e.g. Bach et al. 2014, Piketty 2014). Our paper adds to this literature by assessing the revenue potential of an important policy tool, namely an up-to-date valuation of the property tax base. Furthermore, previous authors have pointed out that the redistributive element of the German property tax in its current form is rather limited (Bach & Schratzenstaller 2013). Our results support this view and indicate that this would not be substantially different once cadastral values are updated. In fact, our findings suggest that the potential for redistribution (if desired by the legislator) depends on the simultaneous reduction of the tax burden on labor.

In addition, our results speak to the literature analysing the distributive effects of tax shifts from labor income towards other tax bases such as consumption (e.g. Pestel & Sommer 2016). So far, little empirical work has been dedicated to property tax related simulations, mostly driven by data limitations. A notable exception is Moscarola et al. (2015), assessing labor market reactions of a property and labor tax reform in Italy. In a similar vein, Figari et al. (2016) investigate the fiscal and distributional consequences of including homeowners' imputed rent in personal taxable income as a kind of property tax for six European countries. Using up-to-date property values to determine property taxes could be regarded as an important complement (and maybe even as a substitute depending on the specific design) to housing income taxation.

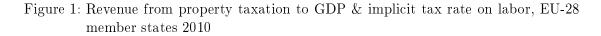
The remainder of this paper is organized as followed: Section 2 illustrates the institutional background of property taxes in Germany. Section 3 describes the matching procedure which combines the two datasets. An analysis of the quality and validity of the matched dataset is provided in Section 4. The simulated tax reform and its distributional and revenue effects are described in Section 5. The final section contains a conclusive discussion of our results.

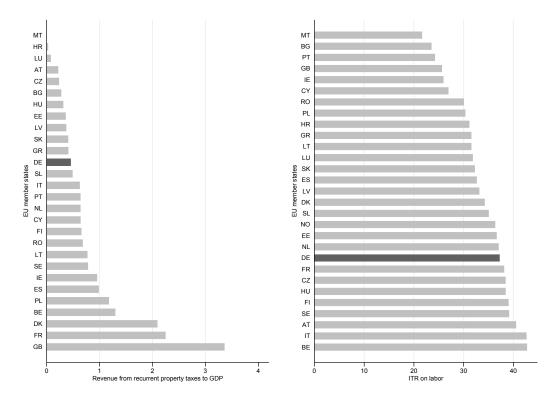
2. Motivation and institutional background

As stressed above, Germany appears to have considerable scope to reform the valuation of property used for property taxation. Basic cross-country comparable descriptives underpin this view. Figure 1 illustrates large disparities across EU-28's member states with regard to revenue of property taxes and the implicit tax rate on labor. Revenues from property taxes are comparatively low for Germany (0.44% of GDP vs. 1.5% in EU-28). At the same time, the ITR on labor in Germany is above average (37.2% vs. 36.1% in EU-28).

So far, several attempts to reform German property taxation have been made, e.g. an overhaul of the *Grundsteuer* was part of the national Reform Program 2014 and 2015 but put on hold hitherto. As a consequence, the current valuation of property dates back to 1964 in Western Germany and to 1935 in Eastern Germany. Back then, rateable values³ were assessed on the basis of capitalized gross returns (i.e. rental income) or, in the case of owner occupied dwellings, on the basis of construction costs (for details

³Please note that we use the term "cadastral value" and "rateable value" interchangeably.





Notes: The left bar chart shows in descending order the percentage of national revenues collected from recurrent property taxes (as % of GDP). The right bar chart compares percentage points of implicit tax rate (ITR) on labor. The ITR is defined as the ratio of all direct and indirect taxes, including social security contributions levied on labor income to total compensation of the employee. *Source:* (European Commission 2013).

see Spahn 2004). The original intention of the legislator was to update the property value on a regular basis, but this was never put into practice.⁴ To make cadastral values comparable, even new buildings, sales or improvements to existing buildings are rated as if they were built several decades ago. Hence, the tax valuation of German properties differ substantially from current market values.⁵ In sum, the link between the property tax liability based on outdated cadastral values and the actual market value of the real estate is very weak (Wissenschaftlicher Beirat BMF 2010).

⁴To partly offset the nominally fixed cadastral values, municipalities apply local tax multipliers in addition to the federal rate. However, the multipliers are set on municipality level only, and hence do not account for heterogenous developments of property values within a given municipality. In addition, the increase in multipliers over time does not match inflation adjustment (see Section 5.1 for details).

⁵Already in 1992 German fiscal authorities executed a comparison of selling prices with underlying cadastral values and found a ratio of ca. 5 to 1 (Bach et al. 2012).

From a policy perspective, two reasons render a reform of the current property tax system in Germany important and hence, our simulation relevant. First, a sunset clause in the German Finanzausgleich - an equalization payment in the German multi-level government - makes its reorganization inevitable by the end of 2018. Since it is often argued that reforms of property tax regimes should be linked to reforms of intergovernmental fiscal frameworks (e.g. Devereux et al. 2007), we consider the sunset clause as a window of opportunity for an overhaul of property taxation in Germany. Second, two pertinent constitutional complaints (BvR 639/11 and 1 BvR 889/12) are currently pending before the Federal Constitutional Court. The court has to decide whether the continued failure to conduct a general reassessment of property values violates the equality-of-treatment clause of the constitution.

3. Description of the data used for the simulation

This paper is based on HFCS and EU-SILC data. The European Union Statistics on Income and Living Conditions (SILC) is a representative survey coordinated by Eurostat that encompasses rich information on income, benefits and taxes, including property taxes paid. Its main limitation is the lack of information on household wealth. In contrast, the Eurosystem Household Finance and Consumption Survey (HFCS) provides detailed data on assets and liabilities, including the (self-assessed) value of real estates households own.⁶ In line with Lindner (2015) and Zhan (2015), we find real estate as the quantitatively most important wealth component of German households. Summary statistics on the two main variables of interest are presented in Table 1. Finally, both surveys contain a number of overlapping variables which we will use below for the matching procedure.

3.1. Methodology of statistical matching

Statistical matching aims to create a dataset from different sources which do not contain the same units. The difference to record linkage, which uses e.g. social security numbers to link *identical* units, is that statistical matching combines *similar* ones (Rässler 2002). Statistical matching in our context allows for imputing the property value Y from HFCS (donor) to SILC (recipient) via a number of appropriate matching variables. These matching variables should be strongly correlated with the merger variable Y and be jointly observed with (Y) as well as (X), i.e. appear in both datasets.

Although EU-SILC does not contain property values, it does provide information on whether a household owns property and how much property tax it pays. Through the

⁶Due to non-response, the most affluent households are likely to be underrepresented in the HFCS. This issue can be addressed by assuming that the upper tail of the wealth distribution approximates a Pareto distribution (Vermeulen 2016). However, this approach is not applicable for subordinate wealth components such as real estate. Importantly, real estate has been found one of the most accurately reported subordinate wealth components in HFCS, with a ratio of reported values in HFCS compared to national accounts amounting to 86% (Eurosystem Household Finance and Consumption Network 2013).

	SILC			HFCS		
	mean	std. dev.	median	mean	std. dev.	- median
tax liability	€ 345	€ 494	€ 250			
main residence				$\in 220,432$	$\in 190,058$	€ 180,000
other property				$\in 264,924$	$\in 678,954$	€ 115,000

Table 1: Summary Statistics of SILC and HFCS

Notes: "Tax liability" stands for the annual property tax liability paid for all owned immovable properties properties. "Main residence" displays the value of the main household residence. "Other property" represents the value of other properties than the main residence.

Source: Own calculations based on sample of property owners in German HFCS and SILC, respectively.

careful selection of matching variables we can assign respondents of EU-SILC (who do own property) the approximate market value of their property. Section A.2 provides a detailed description regarding the choice of appropriate matching variables we use.⁷

We apply a so-called hot deck matching procedure which assigns each observation in HFCS to at least one "nearest neighbor unit" in SILC that is most similar with respect to the matching variables. "Nearest" is defined as the associated observational unit that shows the smallest distance metric based on the set of matching variables. Specifically, we transform the data into uncorrelated, standardized variables with variance equal to 1 and then compute the Euclidean distance between two vectors \underline{x} and \underline{y} (McLachlan 2004). Let C denote covariance matrix and the superscript T the matrix transpose, the distance between a HFCS observation $\underline{x} = (x_1, x_2, x_3, ..., x_N)^T$ and a SILC observation $y = (y_1, y_2, y_3, ..., y_N)^T$ is then defined as:

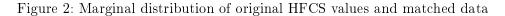
$$d\left(\underline{x},\underline{y}\right) = \sqrt{\left(\underline{x}-\underline{y}\right)^T C^{-1}\left(\underline{x}-\underline{y}\right)} \tag{1}$$

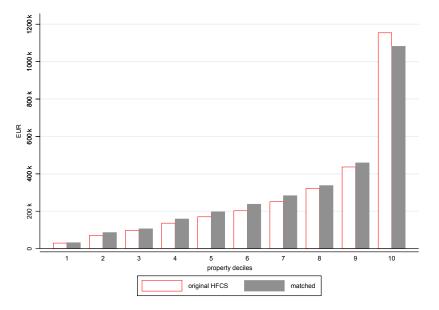
Since our recipient dataset (EU-SILC) is more than three times larger than our donor dataset, donor units may be used for different recipient units repeatedly. Such a marriage algorithm is known as polygamy (Rässler 2002). If the marriage is restricted to a single spouse (monogamy) we would lose almost three quarters of our SILC observations. Hence, we opted for a n > 1 nearest neighbor match with multiple use of donor units (from HFCS). The final matched dataset we generate consists of 13,079 household observations, among which the 6,629 households liable for property taxes are enriched by the market value of their properties. In the next section we will assess the quality of the matched dataset by comparing its properties, marginal, and joint distributions to the original HFCS dataset.

⁷Before applying statistical matching it is important to make sure that the data collection and survey design of HFCS and SILC are comparable. In Section A.1, we discuss coherence requirements such as reference period, target population and collection process of the two datasets.

4. Assessment of the matching result

In order to assess the validity of our matching procedure, we start with analyzing the consistency of the overall marginal distribution. Therefore, we follow established literature and compare the mean value of property owned per property decile between the matched dataset and the original HFCS dataset (Rässler 2002). Visual inspection of Figure 2 shows quite similar distributions of our matched property values. More formally, we perform a two-sample Kolmogorov-Smirnov test comparing the equality of the weighted distributions. Using this test, we cannot reject the null hypothesis that the distribution of property values in the HFCS and the matched dataset are equal.





Notes: The figure displays mean property values per property decile from the original HFCS as well as the matched dataset. The figure is restricted to survey respondents who own property. The white bars represent the distribution of the original HFCS property values. The grey-shaded bars display the distribution of property values in the matched dataset.

As a next step, we analyze the joint distributions of the matching variables and the merger variable in the original HFCS dataset and the matched dataset. Appendix Figure A.4 separately depicts the joint distribution of our merger variable - property value - with each matching variable. Visual inspection of A.4 indicates similar joint distributions in both datasets. Furthermore, we perform parametric tests to detect significant differences in the joint distributions of the matching variables and the merger variable. First, we run two separate univariate regressions of property value on each matching variable. One regression uses the original HFCS dataset while the other regression uses the matched dataset. We then perform a Wald test on the equality of the two estimated coefficients

from both datasets, testing the hypothesis:

 $H_0: coefficient_{HFCS} - coefficient_{matched} = 0$

In order to not only compare means but get a deeper understanding whether the joint distribution is preserved in the matched dataset, the same procedure is conducted using quantile regression. We estimate quantile regressions with coefficients for the 75th quantile.⁸ The first column of Table 2 shows that for the mean regression, the H_0 cannot be rejected across all matching variables. Looking at the results based on quantile regressions (the second column), we continue to find no significant differences in the distribution for most of the matching variables. In sum, our results suggest that both the marginal and joint distributions in the original HFCS are sufficiently preserved in the matched dataset.

	Wald Test		
matching variables	mean regression	.75 quantile regression	
	$\mathrm{Prob}>\mathrm{chi}2$	${ m Prob}>{ m chi2}$	
household size	0.6824	0.6129	
tenure status	0.6037	0.1551	
self-employed	0.9039	0.1218	
total income	0.8655	0.2747	
private pension contribution	0.3472	0.0742	
working hours	0.2727	0.0345	
max age	0.6854	0.5948	
years in residence	0.7232	0.3726	
higher education	0.3201	0.7101	
mortgage dummy	0.3547	0.0520	
years worked	0.4526	0.1632	
rental income dummy	0.9839	0.8015	
public pension income	0.8475	0.1050	

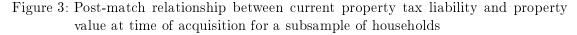
Table 2: Assessing preservation of joint distribution

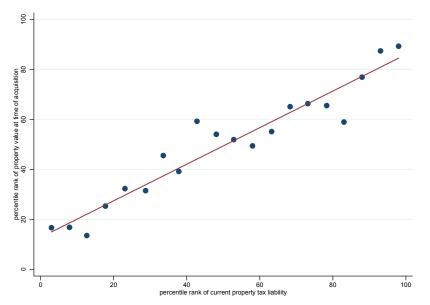
Notes: This table tests whether the coefficients of regressing property value on each matching variable differ between the original HFCS dataset and the matched dataset. Specifically, it displays p-values of a Wald test on the equality of the coefficients. The null hypothesis is that the coefficients obtained from these regressions do not differ between the original HFCS dataset and the matched dataset. The first column shows results from mean regressions, while the second column is based on quantile regressions. See text for details.

As a final step, we make use of auxiliary information to assess the quality and validity of the matched dataset. Specifically, we use the variable *property (market) value at time of acquisition* (which we only observe in HFCS) as an instrument for the current property tax liability (which we only observe in SILC). The idea is that for survey respondents

⁸Using quantile regressions on the .50 and .95 quantile, we receive similar results.

who acquired their property around the year of the last general assessment in 1964, the variable property value at time of acquisition should be highly correlated with the cadastral value of this property, and thus with the current property tax liability. Hence, we can assess the quality of our matched dataset by comparing the (post-match) rank position of the property value at the time of acquisition with the rank position of the current property tax liability of these respondents.⁹ To make this quality assessment valid, we restrict our analysis to households who acquired their property around the year of the last assessment, since the (market) value at the time of acquisition should come very close to the cadastral value of the property (we set a interval of +/-5 years around 1964).¹⁰ Further, we only use households whose only property is their main residence, since property value at time of acquisition is only inquired for the dwelling the household lives in.





Notes: This figure is based on a subsample of households which acquired the property around the year of the last assessment of cadastral values (1964). It presents a binned scatterplot of the relationship between a household's rank position of the property value at time of acquisition and a household's rank position of the current property tax liability using our matched dataset. To construct the figure, we split observations into 20 equal-sized bins based on the rank position of the current property tax liability and plot the mean rank position of the property value at time of acquisition within each bin (the y-axis). See text for details.

Figure 3 presents a binned scatterplot of the mean rank position of the property value at time of acquisition vs. the rank position of the current property tax liability. The

⁹In contrast, for property bought in more recent years, the link between the property value at time of acquisition and its cadastral value is weaker.

¹⁰Please note that in 1964 the legislator defined cadastral values as "the price that could be realized [...] in the case of a sale" (see Spahn 2004.)

rank-rank relationship is almost perfectly linear, suggesting that our matching procedure assigns the underlying property value to the current tax liability reasonably well. The relationship between the two ranks is measured via a Spearman's rho and yields $\rho = 0.74$. Given that we have no information about improvements made to the property since 1964 (which would change the cadastral value of the respective property and hence its property tax liability), we consider this a sufficiently high degree of similarity. In sum, we conclude that our matched dataset should allow for valid inferences, especially on a more aggregated level such as income deciles. In the next section we will run our simulations on this matched dataset.

5. Microsimulation of a property tax reform

Our policy reform simulations are performed on EUROMOD (version G2.0), the taxbenefit microsimulation model designed for EU member states. It applies national taxbenefit policy rules to harmonized micro-data and calculates their effects on household disposable income (Sutherland & Figari 2013). The German component of EUROMOD reproducing the 2010 German tax-benefit system has been validated through comparison with aggregate statistics provided by fiscal authorities (Ochmann & Granados 2011). We run all tax-benefit policy rules at their 2010 setting and then augment the model with a simulated change in property and labor taxation. Hence, our simulation model calculates household disposable income under the current as well as the reformed tax-benefit rules holding everything else constant and, therefore, avoiding endogeneity problems (Bourguignon & Spadaro 2006).¹¹

5.1. Current property taxation and simulated reform

In this section we provide details regarding property taxation in Germany and our proposed policy reform. In our analysis we focus on property taxes levied on (nonagricultural) land, buildings and improvements. All legal regulations of the German property tax, i.e., the definition of the tax base, federal tax rates as well as legal norms regarding the property assessment are set at the federal level. Specifically, the German property tax is calculated as the product of three components: the cadastral value of the property, the federal tax rate and a municipality tax multiplier. Equation (2) formally

¹¹Please note that our simulation abstracts from potential changes in labor supply, avoidance or tax incidence. In general, relatively modest labor supply responses have been found for Germany (e.g. Bargain & Peichl 2017). Furthermore, Löffler & Siegloch (2015) using German data find that in the short run, the incidence of the property tax is borne by landlords. Other scholars argue that this might also be the case in the long-run (Broer 2013). In this context, it is also important to note that two-thirds of the total property tax collected stem from owner-occupied housing, which cannot be shifted onto a third-party. In addition, it has been proposed that a reformed property tax should use legal requirements to prevent shifting of the tax onto tenants (Fuest 2016). Finally, we abstract from administrative costs associated with the proposed policy reform. Bach et al. (2014) estimate the total costs of tax assessment with regard to a one-time capital levy on net wealth for Germany to be modest with around 5%.

shows the calculation of the property tax liability:

(2)

The tax multiplier is set by the local municipality and has been raised by most German municipalities over time (Löffler & Siegloch 2015). This reflects the attempt to at least partly offset the nominally fixed cadastral values. However, using municipality tax multipliers to offset nominally fixed cadastral values does not provide a comprehensive remedy against outdated rateable values. For instance, any adjustment of the tax multiplier occurs on municipality level only, and hence does not account for heterogenous developments of property values within a given municipality.¹²

Federal tax rates have rarely been changed over the last decades and range from 0.26 to 0.35 % for West Germany and from 0.5 to 1 % for East Germany. The main reason why the federal tax rate differs between West and East Germany lies in the different reference year regarding the last assessment of rateable values (1964 for West and 1935 for East Germany, respectively).

Simulated property tax reform. We simulate a property tax reform in which the taxable base - the rateable value - is no more defined by the cadastral value of the property but by its current market value. Since the introduction and rise of the municipality multiplier after 1964 mostly reflects the fact that cadastral values were not adjusted to inflation, we do not apply them when calculating the new property tax liability. This is consistent with the idea to simulate a situation in which current market values (instead of cadastral values) determine property taxes due, which makes the use of inflation-offsetting multipliers redundant. Using current multipliers and current market values would lead to extremely inflated estimates of the new property tax liability. In contrast, not using multipliers when calculating the new property tax liability means that our simulation presents a more conservative estimate of the potential revenue effects of such a reform. Please note that we apply federal tax rates for West Germany to our entire sample, since the reason for the higher federal rate in East Germany is the different reference year regarding the last assessment (1935 instead of 1964), which becomes obsolete when using current market values for all German properties.

Three reform scenarios. We simulate three different scenarios in conjunction with the proposed property tax reform. While the first simulation updates the cadastral values without changing any other taxes, the other two scenarios seek to shift part of the tax burden from labor to property:

- (1) The update of cadastral values is non-revenue neutral: In this first scenario we estimate the additional tax revenue collected from the update of cadastral values irrespective of budget neutrality.
- (2) Revenue neutrality through a lump sum SIC credit: The extra revenue from the update of cadastral values is offset by a non-refundable lump sum SIC credit granted to all employees (all employees with positive SIC).

¹²In addition, the increase in weighted average multipliers since 1974 only accounts for 58% of inflation adjustment (Source: own calculations based on data from the Federal Statistical Office).

• (3) Revenue neutrality through a proportional reduction of employees' SIC: Under this scenario, the additional revenue is used to grant a rebate that is proportional to the SIC payment of an employee.¹³

The first scenario functions as a gauge for the distributive effects from the sole update of cadastral values. The second reform scenario provides a simulation that especially benefits employees at the lower end of the income distribution, where the current tax wedge is particularly large. In the third scenario the size of the SIC rebate is more closely tied to the current SIC payment of the employee.

6. Simulation results

6.1. Revenue effects

We start with the overall revenue effect of the proposed property tax reform. The current annual property tax liability for German households owning property equals \in 345 on average. The proposed property tax reform changing from cadastral values to market values would raise this average property tax liability to \in 967. This would increase the total revenue collected from property taxes substantially from currently \in 5.8 bil. to \in 16.3 bil.¹⁴ In our second scenario (2), we use this additional revenue to grant a credit on SIC in the amount of maximum \in 233 per employee, guaranteeing revenue neutrality. For the average household, this would reduce annual social insurance contributions from \in 6,245 to \in 5,920. In our third scenario (3), we apply the additional revenue to grant a 5.2 % rebate on the SIC payment of every employee, again under revenue neutrality.

6.2. Distributive effects

Now we want to analyze in greater detail how the reform of the property tax and the different scenarios would affect groups of taxpayers differently. Specifically, we examine how the burden of the update of cadastral values is distributed across income deciles of **i**) property-owners only and **ii**) the overall population.¹⁵

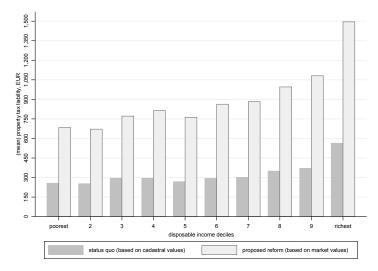
i) We start with examining changes in household budgets following the increase in property tax liability for proprietors only. Figure 4 shows pre- and post-reform property tax liabilities across income deciles of property owners. It is evident from the figure that the increase in the property tax liability is relatively constant across the income distribution of proprietors with an only slightly more pronounced increase for the top five deciles. The post-reform property tax liability for each household income decile is

¹³In the case of joint filers, the simulated tax reduction (i.e. the lump sum SIC credit and SIC rate reduction, respectively) is granted at household level.

¹⁴Please note that our analysis ceteris paribus focuses on property taxes paid by private households excluding corporations.

¹⁵In the following analysis we use equivalized disposable income calculated as market income plus public transfers minus taxes and social insurance contributions. In accordance with established practice, we do account for differences in household size by applying the modified OECD equivalence scale. For details, see OECD (2013).

Figure 4: Pre- and post-reform property tax liability across disposable income deciles (property owners only)



Notes: This figure shows property tax liabilities across disposable income deciles for property owners only. The dark gray bars display the mean pre-reform property tax liability. The light gray bars show the mean post-reform property tax liability.

Source: Own calculations based on EUROMOD.

approximately three times larger, compared to a pre-reform situation. Hence, the relative size of the property tax liability across the income distribution of homeowners is by and large preserved under the proposed reform. Put differently, an update of cadastral values does not hurt a certain income group of property owners in particular.

ii) Next we want to study the effect of the proposed update of cadastral values across the entire income distribution (regardless of being homeowner or not). We start with scenario (1), which is the non-revenue neutral simulation. The bars in Figure 5 show the change in disposable income in absolute monetary values (EUR) by disposable income decile. The negative change in income increases with household income, which is expected given that ownership rates in Germany rise substantially with income (see Appendix A.5).

When displaying the relative income change, a different picture emerges. The triangles in Figure 5, representing the percentage change in disposable income under the reform scenario (1), vary little across the distribution.¹⁶ Hence, poorer households are in relative terms as affected by the update of cadastral values as richer ones. Thus, an update of cadastral values without redistributing the additional revenue would render such a reform virtually neutral in terms of redistribution. Previous authors have noted that the redistributive element of the German property tax is limited (e.g. Bach & Schratzenstaller

¹⁶The second decile stands out with a similar absolute increase in tax burden of proprietors and a only slightly higher ownership rate, but a considerable higher income than the first decile.

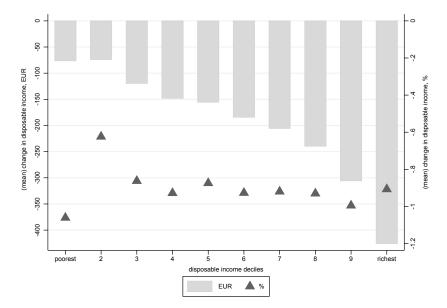


Figure 5: Change in average household disposable income per decile under scenario (1)

Notes: This figure displays the change in disposable income under reform scenario (1), which is the non-revenue neutral simulation. The figure is based on the entire population, regardless of being homeowner or not. Households are put into deciles according to their pre-reform disposable income. Bars show the change in disposable income in monetary values (EUR); triangles display the percentage change in disposable income. *Source:* Own calculations based on EUROMOD.

2013, Broer 2013). Our findings suggest that this would not change when cadastral values are updated. This motivates the analysis of our revenue-neutral reform scenario (2) and (3), where the additional tax revenue is used to lower the tax burden on labor.

Scenario (2) is a simulation in which the additional tax revenue of the proposed update of cadastral values is offset by a non-refundable lump sum SIC credit. Such a lump sum SIC credit corresponds with a relative high tax relief for low income earners, whose contribution rate is reduced to a relatively greater extent. Figure 6 displays the income change on deciles of household disposable income under reform scenario (2). The figure shows that all household deciles would gain in disposable income except of the top three ones.¹⁷ The total yearly gains range between $\in 20$ and $\in 60$ on average across the household income deciles, whereas the three top deciles loose between $\in 15$ and $\in 209$ on average. The triangles in Figure 6 display the income change relative to disposable household income, ranging between +0.43 % and -0.40 % for the single deciles.

As a next step we turn to our third reform scenario (3), in which the additional revenue is used to grant a rebate that is proportional to the SIC payment of an employee. Specifically, we simulate a 5.2 % rebate on the social insurance contribution paid by the employee. The impetus for scenario (3) is that employees should enjoy a proportional

¹⁷The reason why the first decile gains relatively little has to do with the fact that households in this decile are disproportionately more likely to be unemployed, out-of-labor force, or self-employed, which corresponds with zero SIC payments. The same holds for the first decile of scenario (3).

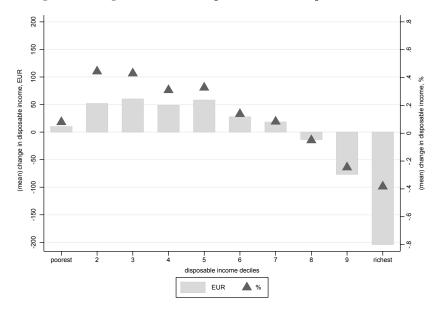


Figure 6: Change in average household disposable income per decile under scenario (2)

Notes: This figure displays the change in disposable income across deciles of disposable income under reform scenario (2), which is the lump sum SIC credit simulation. The figure is based on the entire population, regardless of being homeowner or not. Households are put into deciles according to their pre-reform disposable income. Bars show the disposable income change in absolute monetary values (EUR); triangles display the percentage change in disposable income.

Source: Own calculations based on EUROMOD.

reduction of their SIC payments. Figure 7 displays the income change relative to deciles of disposable household income under reform scenario (3). The figure indicates that the proportional rebate would have only small effects in terms of redistribution. With exception of the first decile, which clearly suffers, the average losses and gains per income decile do not exceed 0.2 % of income. Similarly, absolute changes in disposable income across income deciles do not exceed $\in 50$. In sum, it seems that middle-income households would profit to some extent from this reform scenario, whereas low- and high-income households slightly suffer.

Figure 8 provides additional insights into the distributional effects of our simulations. For each of our two revenue-neutral scenarios we now display the share of gainers and losers per disposable income decile. A household is defined as a gainer (loser) when experiencing a positive (negative) change in disposable income. The upper part of Figure 8 shows the result for reform scenario (2). We find more gainers than losers, with the share of losers increasing steadily with the income level. In contrast, the share of gainers is much more evenly distributed across income deciles. Turning towards scenario (3), we find again more gainers than losers, but this time losers are less concentrated in the upper part of the income distribution than under scenario (2). This mirrors our results of Figure 7, suggesting that the proportional rebate would have only small effects regarding the income distribution. Please note the share of gainers generally exceeds the share of

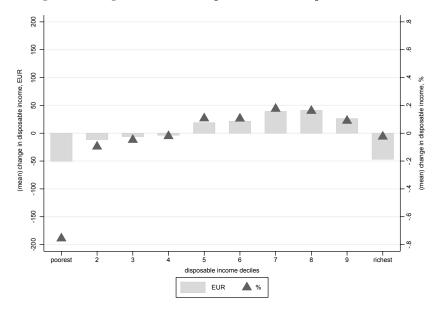


Figure 7: Change in average household disposable income per decile under scenario (3)

Notes: This figure displays the change in disposable income under reform scenario (3), which is the proportional rebate simulation. The figure is based on the entire population, regardless of being homeowner or not. Households are put into deciles according to their pre-reform disposable income. Bars show the disposable income change in absolute monetary values (EUR). Triangles display the percentage change in disposable income. *Source:* Own calculations based on EUROMOD.

losers across all income deciles, except for the top income decile under scenario (2). In contrast, the mean change in disposable income is negative for three income deciles under scenario (2) (see Figure 6) and for five income deciles under scenario (3) (see Figure 7). Thus, we conclude that gains of the tax shift are modest but widespread, whereas losses tend to be bigger but less frequent.

Finally, we want to assess overall changes in inequality associated with our three reform scenarios. For this purpose, we employ two widely used inequality indices, namely the Gini and the Atkinson with $A_{\varepsilon} = 1$. In line with our previous results, we find the nonrevenue neutral scenario (1) to barely change the distribution of income (see Table 3). Regarding scenario (2), we observe a small reduction in income inequality. In contrast, scenario (3) would widen the income distribution, though only very slightly. Looking at changes in poverty thresholds (set at 60 % of median disposable income), we barely find any effect of the three reform scenarios. However, this does not rule out that the proposed tax shift may generate significant gainers and losers. As the comparison of extensive margin (see Figure 8) with the intensive margin (see Figures 6 and 7) already suggests, the worst off 1% might be affected by a considerable income shock in both scenarios (2) and (3). These scenarios could therefore potentially face opposition from asset-rich but income-poor households, which might ask for mitigating measures. This seems to be pertinent especially to the political acceptance of such a reform, given that the issue of property taxation can affect election outcomes (Bosch & Solé-Ollé 2007).

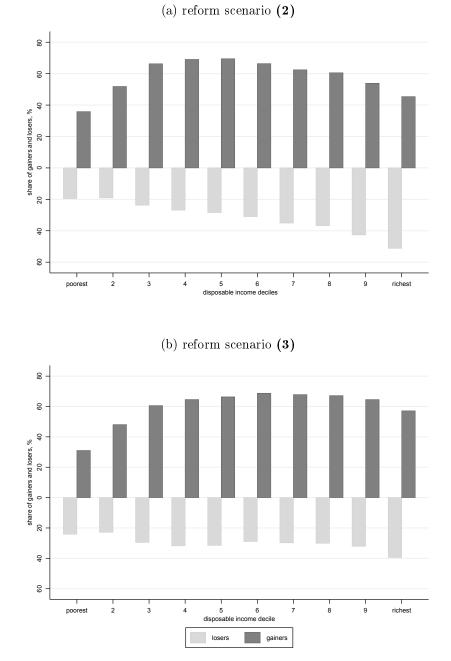


Figure 8: Shares of gainers and losers per disposable income decile under reform scenarios (2) and (3)

Notes: This figure shows the share of gainers and losers for the reform scenarios (2) and (3). Gainers (losers) are defined as households with a positive (negative) change in disposable income under the respective reform scenario. *Source:* Own calculations based on EUROMOD.

LL N		Status Quo	Non-revenue neutral (1)	SIC lump sum credit (2)	Proportional SIC rebate (3)
Budgetary	private property tax revenue	€ 5.8 bil.	€ 16.3 bil.	€ 16.3 bil.	€ 16.3 bil.
$\mathbf{B}\mathbf{u}$	ITR on labor	37.2~%	37.2~%	36.5~%	36.5~%
ve	Gini	0.2677	0.2678	0.2665	0.2683
Distributive	Atkinson (1)	0.1147	0.1148	0.1137	0.1152
rib	Poverty threshold	€ 950	€ 936	€ 950	949 €
ist	worst off 1 $\%$			- 13.6 $\%$ income	- 14.4 $\%$ income
	best off 1 $\%$			+ 4.5 $%$ income	+ 2.6 $%$ income

Table 3: Revenue and distributive effects of the three scenarios

Notes: "Status Quo" represents the pre-reform situation. "Private property tax revenue" depicts property tax revenues collected from private households, "worst/best off 1 %" depicts the mean percentage loss/gain in disposable income of the 1 % most affected households. *Source:* Own calculations based on EUROMOD.

7. Conclusion

The idea of higher taxes on land, capital and wealth to finance mounting public debt has gained ground in several OECD countries. At the same time, the scope for shifting taxes to more growth-friendly revenue sources appears underused in many European countries. This seems to be especially true for Germany, a country which makes only little use of property taxes while at the same time having a high implicit tax rate on labor. Against this backdrop we simulate a property tax reform for Germany which increases revenues from the taxation of property while simultaneously lowering the tax burden on labor. Changing the current property tax scheme based on outdated cadastral values to one based on market property values, we find substantial revenue effects of the proposed reform. Specifically, tax collection from private household property would increase from currently \in 5.8 bil. to \in 16.3 bil., allowing for an overall reduction of the implicit tax rate on labor from 37.2% to 36.5%. Using EU-28 cross-country levels as a comparison, this equates in an improvement of the implicit tax rate on labor by three positions. In contrast, the increase in the ratio of property tax revenue to GDP would change Germany's position by 13 places, with an after-reform level similar to Denmark's (compare Figure 1). Examining the distributional effects of the reform on household level, we find the update of cadastral values without using the additional revenue to lower SIC to be virtually neutral in terms of redistribution. As rich and poor households show comparable increases in the (relative) property tax burden, any potential redistribution under the proposed reform depends crucially on the design of the revenue-neutral SIC reduction. While a SIC reduction via a lump-sum tax credit would especially benefit low-income households, a SIC rebate proportional to households' current contributions would barely alter the overall distribution of disposable income. This gives policy-makers

considerable scope via the specific design of such a reform.

In light of the controversial nature of the outdated taxation of property in Germany and the apparent reluctance of policy-makers to tackle it, our paper reduces uncertainty about both revenue and distributional effects of such a reform. Depending on the exact design, our results suggest that low- and median income households could be made better off when reducing the overall tax burden on labor.

We are aware that shifting taxes from labor to property is not easy to implement, especially in a federal system like Germany where property taxes accrue to local municipalities, and social insurance contributions to federal budgets. In addition, mass appraisal can be both expensive and seem as intrusive. However, our analysis aims to inform about the fiscal and distributional effects of such a shift, which can then be mapped against institutional costs and legal constraints. While such an analysis is beyond the scope of our paper, it provides a fruitful avenue for future research.

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A. Appendix

The following sections provide a detailed description of the different steps taken to construct the matched dataset.

A.1. Coherence check of datasets

Before applying statistical matching, it is important to make sure that data collection and survey design of HFCS and SILC are comparable. As a matter of fact, HFCS and EU-SILC have the same target reference population, namely all private households in Germany. Both surveys exclude all institutionalized population, i.e. people living in retirement homes, health care, religious, correctional and penal institutions. The reference units are defined as all age 16+ members currently living in the same household. Reference point for balance sheet items is in both surveys the date of interview. Interviews for EU-SILC were held between 05/2010 - 11/2010. The field work for HFCS data was conducted from 09/2010 - 07/2011. Both surveys use the same income reference period, which is 2009. Finally, due to the potential non-response bias, HFCS tries to oversample wealthier households. In contrast, SILC does not apply such oversampling. However, it is important to note that for Germany, we find one of the best data coherence between HFCS and EU-SILC among the 15 euro area countries regarding potential matching variables. For instance, the median annual gross income differs by less than \in 100. This small difference in median annual gross income despite oversampling might reflect the very limited oversampling of HFCS in Germany. Oversampling in Germany was only based on geographic information about the distribution of taxable income, whereas other countries applied much more rigorous oversampling based on, e.g., wealth tax records. In sum, we conclude that regarding target population, household definition and reference period, the two survey designs appear to be sufficiently coherent to allow statistical matching.

A.2. Identification of matching variables

As mentioned in the main text, the careful selection of the matching variables is crucial when using statistical matching (Little & Rubin 2014). In the spirit of the stepwise approach of Leulescu & Agafitei (2013), we apply the following three key steps to choose appropriate matching variables:

First, we carry out a data reconciliation process to correct variable discrepancies of HFCS and SILC due to the use of different technical definitions or variable concepts. For instance, we harmonize potential matching variables when their scale of measure differs. Sometimes such harmonization is not possible when the level of detail and accuracy lie far apart. In such cases, we do not consider these variables for the matching procedure.¹⁸

¹⁸To give an example, HFCS inquired total welfare benefit transfers on household level, whereas SILC collected at an individual level old-age benefits, survivor benefits, sickness pay, disability benefits and education-related allowances separately. As a consequence, we had to eliminate gross cash income from regular social transfer from the potential set of matching variables.

A.3. Reconciliation process of common variables - HFCS 2010 and SILC 2010

Raw Data	Recoded Household Data	Coarsened	Variable
	Average age in household		average age
Month and year of birth	Age of oldest household member	- In SILC	max age
	Number of females per household		number of females
Gender, age	Number of children per household	In SILC	number of children
	Absolute household size		household size
Relationship status	Household's family structure	In HFCS	marital status
Country of birth, citizenship	Dummy=1 if any household member with migration background	In HFCS	any migrant
Duration of stay in host country	Maximum number of years spent in Germany by any household member	In HFCS	years in country
Self-defined economic status	Dummy=1 if at least two persons are active on labor market (one if single-person -household or lone)	In HFCS	employment status
Highest ISCED level attained	Dummy $=1$ if any household member has a higher education degree	In SILC	higher education
Occupation (ISCO-88)	Dummy for each ISCO group=1 if any household member' occupation fits to the respective group	In SILC	ISCO-88
Economic activity (NACE)	Dummy for each NACE group=1 if any household member' economic activity fits to the respective group	In SILC	NACE
Type of labor contract	Dummy =1 if all active household members have a permanent contract	In SILC	work contract
Labor force experience	Maximum number of years spent in paid work by any household member	No coarsen	years worked
Number of hours worked per week	Maximum number of hours worked by any household member	No coarsen	working hours
Individual gross employee income	Aggregated household's cash employee income	No coarsen	employee income
Individual gross self-	Aggregated household's cash self- employed income	— No coarsen	self-employed income
employed income	Dummy=1 if any household member receives self_employed income	— No coarsen	self-employed dumm
Individual income from public pension	Aggregated household's income from public pension scheme	No coarsen	public pension income
Individual income from unemployment benefit	Aggregated household's income from unemployment benefit	No coarsen	unemployment benefit
Income from financial investment	Originally on household level	-	income from financia

to be continued

Raw data	Recoded Household Data	Coarsened	Variable	
Income from other public	Aggregated household's income from other public	No coarsen	other benefits	
transfers	transfers	No coarsen		
Contribution to private	Aggregated household's contribution to private pension	No coarsen	private pension	
pension plan	plans	No coarsen		
			contribution	
Household main residence	Originally on household level	- in HFCS	mortgage	
mortgage principle payment,	Dummy=1 if there ar any mortages using household		mortgage dummy	
mortgage interest rate	main residence as collateral			
Bental income	Originally on household level		rental income	
Rental income	Dummy=1 if any household member receives rental			
	income		income rental dummy	
Intra-household cash transfer	Originally on household level	-	intra-household	
			income	
Total gross income	Originally on household level	-	total income	
Tenure status	Originally on household level	-	tenure status	
Rental payment	Originally on household level	-	rent	
Years living in main			,	
household residence	Originally on household level	-	years in residence	
Possession of vehicle	Dummy=1 if any household member owns a car	-	vehicle	
Income from regular social				
transfer	-	Incoherent	-	
Non-collateralized debt	-	Incoherent	-	
Other self-employed income	-	Incoherent	-	

Table A.3 provides a comprehensive summary of the reconciliation process and a list of the common set of variables from both surveys.

Second, it is important that the common set of variables (i.e. our potential matching variables), which appear both in HFCS and SILC, show similar distributions. We apply Hellinger Distance (HD), a measure to evaluate similarity of variable distribution of two different datasets (Webber & Tonkin 2013, Eurostat 2013). Equation (3) assesses the similarity/dissimilarity between donor HFCS and recipient SILC for each potential matching variable. A HD value of 0 can be interpreted as perfect similar and a value of 1 as perfect discrepancy. As commonly stated in the literature, an HD of over 5% raises concerns about the similarity in marginal distributions (e.g., Leulescu & Agafitei 2013).

$$HD\left(V,V'\right) = \sqrt{\sum_{n}^{K} \left(\sqrt{\frac{n_{Di}}{N_{D}}} - \sqrt{\frac{n_{Ri}}{N_{R}}}\right)^{2}} \tag{3}$$

V is the donor dataset (HFCS) and V' the recipient dataset (SILC), K is the total number of cells in a contingency table, n_{Di} is the frequency of cell *i* in donor data D, n_{Ri} is the frequency of cell *i* in recipient data R, and N is the total size of the specific contingency table.

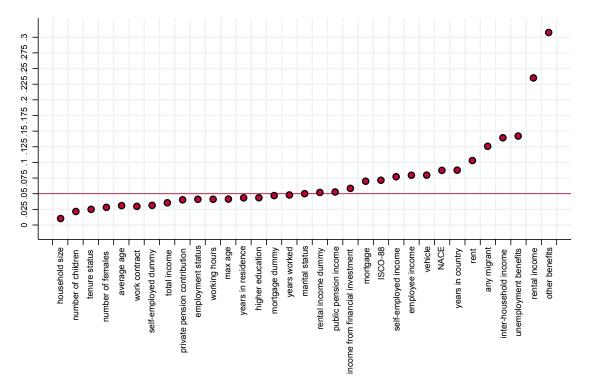


Figure A. 1: Comparing extensive margins of common variables

Notes: Hellinger Distance metrics for common coherent variables with reference line at 0,05. For further information on variables abbreviations, we refer to appendix A.3. *Source:* Own calculations based on sample of property owners in German HFCS and SILC, respectively.

We calculate HD metrics on a truncated dataset. To be more precise, only HFCS units that own property are taken into account as only this subsample is liable to property taxation. This restriction reduces noise as it prevents the matching of property values to households not liable for property taxation. In a similar vein, we also restrict the recipient file to observations liable to property taxes.

Figure A.1 indicates that for quite some variables, the HD metric is below 5%. For instance, most of the demographic variables from both surveys show a strong degree of similarity regarding their distributions. Furthermore, total household (gross) income and contributions to private pension plans are very similar across both surveys. More importantly, variables capturing whether a person has rental income or tenure status are very evenly distributed in both surveys. Unsurprisingly, relatively low similarity is found for variables measuring welfare transfers. All other variables exceeding the 5% threshold are not used for the matching, as this would introduce noise to our analysis. Additional tests comparing weighted means by using simple t-tests confirm our selection of suitable variables based on the HD metrics (results available upon request).

As a third step, we want to test the explanatory power of the set of common variables which fulfill the condition of coherence and similarity of distributions (i.e. all variables not exceeding the 5% threshold in Figure A.1). According to D'Orazio et al. (2006), common variables for matching should be selected on the basis that they significantly explain the variation in the merger variable Y, that is the value of properties owned. As standard in the literature, the null hypothesis of no association between common variables and market value of property is tested. We run Rao-Scott tests, a correction of Chi-squared tests for contingency tables when the estimated cell proportions are derived from survey data (Rao & Scott 1981). In order to also provide a measure of strength of association between two variables, the Pearson correlation coefficients are calculated. Table A.2 shows results for the Rao-Scott test and Pearson correlation coefficient.¹⁹ As depicted, 13 of the 19 variables that have been found to be similarly distributed across both surveys are also significantly correlated with our merger variable Y. When regressing the market value of property owned (= Y) on such 13 variables, we obtain a R^2 of 0.64. Hence, based on overall coherence, similar distributions and sufficient predictive power, we select these 13 variables for statistical matching. Table A.3 provides an overview of all variables considered for statistical matching, with the 13 variables finally selected for statistical matching shaded in gray.

¹⁹Our results stay qualitatively the same when applying multivariate statistics such as stepwise regressions (results available upon request).

Common Variables	Test of Independence		Pairwise Correlatio	
	Pearson	0.0481	0.05048	
max age	Likelihood ratio	0.0451	0.0504*	
	Pearson	0.0951	0.00.41	
average age	Likelihood ratio	0.1093	0.0641	
	Pearson	0.0353	0.0533	
number of females	Likelihood ratio	0.0224	0.0533	
	Pearson	0.2612	0.0000	
number of children	Likelihood ratio	0.2859	-0.0238	
	Pearson	0.0029	1	
household size	Likelihood ratio	0.0016	0.0376*	
	Pearson	0.0897		
marital status	Likelihood ratio	0.08765	0.0308	
1 4 4 5	Pearson	0,1281	0.0007	
employment status	Likelihood ratio	0,1145	0.0967	
	Pearson	0.2731		
higher education	Likelihood ratio	0.2659	0.1473**	
	Pearson	0.0000		
work contract	Likelihood ratio	0.0000	$0,\!1154$	
	Pearson	0.0962	0.000	
years worked	Likelihood ratio	0.1496	0.0665*	
	Pearson	0.0004		
working hours	Likelihood ratio	.0.0000	0.0568*	
	Pearson	0.0003		
self-employed dummy	Likelihood ratio	0.0003	0.1545* *	
	Pearson	0.04557		
public pension income	Likelihood ratio	0.0180	0.0951*	
	Pearson	0.0000		
private pension contribution	Likelihood ratio	0.0000	0.1092*	
	Pearson	0.0033		
mortgage dummy	Likelihood ratio	0.0027	- 0.0513*	
	Pearson	0.0000		
rental income dummy	Likelihood ratio	0.0000	0.2975**	
	Pearson	0.0000		
total income	Likelihood ratio	- 0.0000	0.3296**	
	Pearson	0.0000		
tenure status	Likelihood ratio	0.0000	0.0280^{*}	
	Pearson	0.0311		
years in residence	 Likelihood ratio	0.0304	0.0068*	

Table A. 2: Rao Scott Test and Pearson Correlation Coefficients common variables

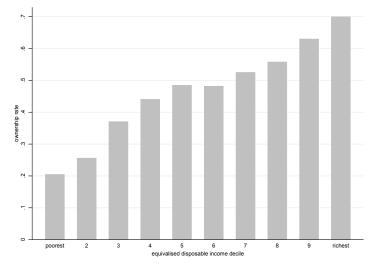
Notes: Tests of independence - dichotomized for continous variables - cover Pearson's and likelihood-ratio chi-squared, both corrected for the survey design with the second-order correction of Rao and Scott (1984). Pairwise correlation coefficients are calculated allowing for sample design. Significance levels are based on survey-based variance estimates, with * and ** indicating significance at 5- and 1-percent levels, respectively.

Variable Name	Coherence	Distributional Similarity	Explanatory Power
max age	~	✓	1
average age	~	×	×
number of females	~	×	×
number of children	~	×	×
household size	1	1	1
marital status	~	V	×
any migrant	~	×	
years in country	~	×	
employment status	~	×	×
higher education	1	V	1
ISCO-88	~	×	
NACE	~	×	
work contract	~	~	×
years worked	1	v	1
working hours	1	~	1
employee income	~	×	
self-employed income	~	×	
self-employed dummy	1	V	1
public pension income	1	1	1
unemployment benefit	~	×	
income from financial investment	~	×	
other benefits	~	×	
private pension contribution	1	v	1
mortgage	1	×	
mortgage dummy	~	1	1
rental income	1	×	
rental income dummy	1	1	1
intra-household income	1	×	
total income	~	1	1
tenure status	1	1	1
rent	1	×	
years in residence	1	1	1
vehicle	1	×	

Table A. 3: Summary of matching variables



A.4. Joint distribution of merger variable with each matching variable for original HFCS values and matched results from three approaches



A.5. Ownership rates across disposable income deciles

Notes: Ownership rate comprises main household residence and/or other properties. Source: Own calculations based on SILC.