

Kraft, Kornelius; Lammers, Alexander

Conference Paper

Bargaining Power and the Labor Share - a Structural Break Approach

Beiträge zur Jahrestagung des Vereins für Socialpolitik 2021: Climate Economics

Provided in Cooperation with:

Verein für Socialpolitik / German Economic Association

Suggested Citation: Kraft, Kornelius; Lammers, Alexander (2021) : Bargaining Power and the Labor Share - a Structural Break Approach, Beiträge zur Jahrestagung des Vereins für Socialpolitik 2021: Climate Economics, ZBW - Leibniz Information Centre for Economics, Kiel, Hamburg

This Version is available at:

<https://hdl.handle.net/10419/242342>

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.

Bargaining Power and the Labor Share – a Structural Break Approach

KORNELIUS KRAFT[†] ALEXANDER LAMMERS[‡]

August 2021

Abstract

In this paper we investigate the relevance of bargaining institutions for the decline of the labor share. Several explanations for the decline exist, which consider the relevance of technology, globalization and markups. Neglected so far is the influence of bargaining institutions, in particular with a focus on changes in the outside option. We provide evidence of this issue, using the Hartz IV labor market reform in Germany as an exogenous shock in the wage bargaining of employees, and investigate its impact on the labor share. We begin by developing a theoretical model in which we outline the effect of a decrease in the outside option within a wage bargaining framework. Thereafter, the approach is twofold. Combining the EU KLEMS and Penn World Table databases, we first endogenously identify the Hartz IV reform as a significant structural break in the German labor share. Second, we estimate the effect of the Hartz IV legislation on the aggregated labor share using a synthetic control approach in which we construct a counterfactual Germany doppehganger. Finally, we use rich firm-level panel data compiled by Bureau van Dijk to support our results on the aggregated labor share. We find that the reform decreases the labor share by 1.6 – 2.7 percentage points depending on method and aggregation level. The synthetic control approach furthermore provides evidence that this effect is persistent over time since the reform.

Keywords: Labor share, Hartz reform, change-point tests, inequality, synthetic control method, GMM

JEL Codes: D04, E25, J51, L21, O43

Acknowledgments: We thank the German Research Foundation (DFG) for its financial support of the Collaborative Research Center “Statistical modeling of non-linear dynamic processes” (SFB 823, Project A5). We received very helpful comments and suggestions from the SFB 823 “Structural Break Day” in Bochum, the annual meeting of the SFB 823 in Lüdenscheid, the internal colloquium of the chair of Economic Policy, the (virtual) 13th IAAEU Workshop on Labour Economics and the (virtual) conference of the Hans-Böckler-Foundation. Any remaining mistakes are our own.

[†]TU Dortmund University, Faculty of Business and Economics, Vogelpothsweg 87, D-44227 Dortmund, ZEW Mannheim, IZA Bonn and KU Leuven, Belgium. Email: Kornelius.Kraft@tu-dortmund.de.

[‡]Corresponding author: TU Dortmund University, Faculty of Business and Economics, Vogelpothsweg 87, D-44227 Dortmund. Email: Alexander.Lammers@tu-dortmund.de.

1 Introduction

There is intense debate on how economic outputs are divided between capital and labor (Rodriguez & Jayadev 2013). Usually, this division is measured using the concept of the labor share which is the ratio of labor compensation to economic output. In macroeconomic models, the stability of the labor share is often referred to as a stylized fact of growth (e.g. Kaldor 1957). This stability, however, is challenged since declining labor shares are observed in many countries over several decades (e.g. Barkai 2020; Cantore et al. 2020; Karabarbounis & Neiman 2014).¹ In Germany for example, the labor share declined by roughly five percentage points from 70 % in the 1970s to 65 % in 2015 which also raises distributional questions regarding inequality (e.g. Iñaki 2020; Card et al. 2020; Piketty & Zucman 2014).

Regarding potential determinants on the decline of labor shares, there is a wealth of theoretical and empirical literature. One explanation named in this context is technological progress such as the use of robots and algorithms as well as a decreasing price of capital in relation to labor.² Another line of research emphasized the role of so-called ‘superstar firms’. These firms are based on capital-intensive production and exponential growth.³ Globalization combined with outsourcing of labor-intensive tasks is another explanation.⁴

In this paper we consider the decline of bargaining power as an so far overlooked reason for the observed decrease in labor share, in particular with a focus on changes in the outside option. We therefore utilize the unique exogenous reform shock of the Hartz IV legislation, leading to a decrease in the threat point of unions within a bargaining framework. For the investigation, our approach is twofold. We first show that the Hartz legislation and, in particular, the Hartz IV reform in Germany contribute

¹Figure A.1 in the Appendix provides an overview of developments for different countries.

²See for example the literature by Acemoglu et al. (2020); Eden & Gaggl (2018); Acemoglu & Restrepo (2018); Acemoglu (2003); Bentolila & Saint-Paul (2003). Results from these studies suggest that the labor share declined by 4 to 6.3 percentage points for firms that adopt robots (Acemoglu et al. 2020).

³This strand of the literature is in particular driven by work from Autor et al. (2020); De Loecker et al. (2020); Kehrig & Vincent (2021).

⁴See for example Elsby et al. (2013) in the context of offshoring and Stockhammer (2017) for the impact of financial globalization.

to a significant structural break in the time series of the aggregated labor share. We therefore apply several endogenous tests drawn from the change point literature (e.g. [Antoch et al. 2019](#); [Andrews 2003](#); [Bai & Perron 2003](#)) in which we identify the Hartz IV reform as a significant structural break. In a second step, we estimate the reform effect on the labor share using (i) data on the aggregated labor share and (ii) firm-level data (i.e. the ‘dafne’ dataset) compiled by Bureau van Dijk. We apply ordinary least squares as well as a synthetic control approach (e.g. [Abadie 2021](#)) in which we construct a Germany doppelganger as a counterfactual for what would have happened with the German labor share in the absence of the Hartz IV reform. For this analysis we use the EU KLEMS data combined with the Penn World Table database to investigate the German labor share for the period 1970 to 2015. Regarding the firm-level data, we apply fixed effects as well as System GMM estimation techniques. We provide evidence that the exogenous shock of the Hartz IV reform reduces the German labor share by around 2 percentage points. The synthetic approach additionally suggests that this decline is lasting, at least up to ten years after the reform was implemented.

Related literature exists which examines the relationship between bargaining power and the labor share, however with a different focus. For the aggregate labor share, [Young & Zuleta \(2018\)](#); [Blanchard & Giavazzi \(2003\)](#) consider the direct bargaining power of unions. In a similar vein, [Fichtenbaum \(2011\)](#) finds that the decline in union density explains roughly one third of the decline in the share of labor. [Bental & Demougin \(2010\)](#) develop a theoretical model which explain movements in the labor share which depend on labor market institutions, and [Brock & Dobbelaere \(2006\)](#) develop a bargaining framework for the effects of globalization on the labor share. More closely related are, for example, [Bazillier & Najman \(2017\)](#), who investigate how crisis events affect the threat points of workers. More recently, [Stansbury & Summers \(2020\)](#) investigate the relevance of bargaining institutions for workers in the United States and [Ciminelli et al. \(2020\)](#) consider the impact of job protection deregulation. These authors in particular find that the decline in workers’ bargaining power might be the main reason for changes in the labor share. On the firm level, there are numerous stud-

ies examining the role of firm-specific factors such as workforce or firm characteristics for the labor share (Harju et al. 2021; Siegenthaler & Stucki 2015). There is, however, so far no study on legislative action which renders the outside option in a bargaining context less attractive.

The remainder of this paper proceeds as follows. In Section 2 we provide the institutional framework and background of the Hartz legislation in Germany. Moreover, we provide a simple bargaining model in which we derive implications for the connection between changes in the outside option in wage bargaining and the labor share. Section 3 then provides empirical evidence on the Hartz IV reform and identifies the reform as a structural break in the German labor share where we apply several endogenous and exogenous change point tests. Section 4 provides estimates of the magnitude of the reform on the aggregated labor share and Section 5 provides firm-level evidence. Section 6 concludes and provides policy implications.

2 Institutional framework in Germany

2.1 The Hartz legislation

A persistent and high unemployment rate in Germany in the early years of the 21st century led to the implementation of the so-called Hartz reforms, named after the chairman of the commission, Peter Hartz.⁵ The reform consisted of four packages (Hartz I–Hartz IV), implemented successively during the years 2003–2005 and were designed to increase the flexibility of the labor market. Their main purpose was to reduce long-term unemployment.

The reform package started with Hartz I and II, introduced on January 1st, 2003. Both of these reforms led to increased labor market flexibility by deregulating temporary work, dismissals and fixed-term contracts. Empirical evidence on these two reforms is provided, for example, by Bradley & Kügler (2019) who find an increase

⁵The unemployment rate in Germany was persistently high at roughly 10 percent and a peak was reached at 11.1 percent in the year 2005 (Dustmann et al. 2014).

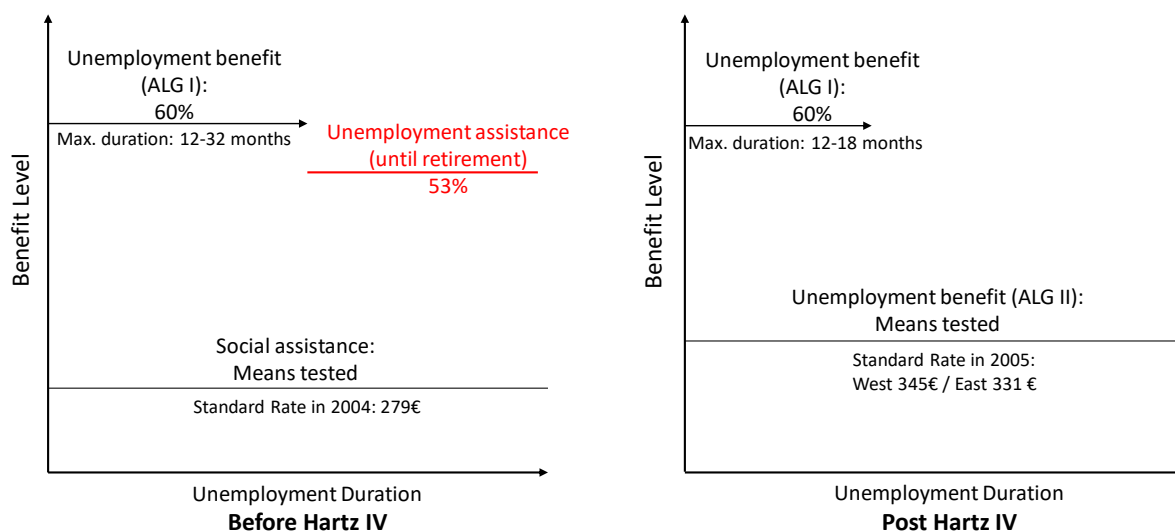
in mini-job usage. The Hartz III reform was aimed at increasing matching efficiency on the labor market by restructuring the Federal Employment Agency. It became effective on January 1st, 2004. Empirical evidence from a macroeconomic point of view on effects of the reform on matching efficiency is, for example, provided by [Launov & Wälde \(2016\)](#) and [Klinger & Weber \(2016\)](#).

Finally, the Fourth Act for Modern Labour Market Services (commonly known as Hartz IV) focused on the abolition of long-term wage-dependent support payments and a transition to fixed benefit levels equivalent to the socio-cultural subsistence level. This final (and centerpiece) part of the reform became effective on January 1st, 2005. Before this reform, there had been a three-tier system consisting of short-term unemployment benefits (*Arbeitslosengeld ALG I*), unemployment assistance (*Arbeitslosenhilfe*) and social assistance (*Sozialhilfe*). The short-term unemployment benefits amounted to roughly 60-67 % of previous earnings and were usually paid for 12 months.

The Hartz IV reform transformed this system into a two-tier system, as depicted in [Figure 1](#). In particular, the reform comprised the following elements: merging of unemployment assistance (*Arbeitslosenhilfe*) and social assistance (*Sozialhilfe*) into unemployment benefit II (*ALG II*); reduction of the period of entitlement to unemployment benefit (ALG I) from a maximum of 32 to a maximum of 18 months; reduction of support for children and young people, extended means testing (crediting of own fortune and income of partners against transfer payments) and new and stricter sanctions for unfulfilled conditions in the search for employment. The reform therefore led to a dramatic cut in received benefits for the long-term unemployed, since they were no longer eligible for long-term unemployment assistance (which was wage-dependent). As a result of the reform the consequences of unemployment in terms of wage cuts were more severe and therefore pressure on the employed to make wage concessions and on the unemployed to accept unattractive job offers increased.

In a similar vein, [Hartung et al. \(2018\)](#) also highlight, particularly for long-term employed workers with previously high wage payments, that the Hartz IV reform repre-

Figure 1: Hartz IV reform: reduction in the outside option



Notes: This figures shows the effects of the Hartz IV reform for a single household. The reform transformed the three-tier system of unemployment benefits ('Arbeitslosengeld'), unemployment assistance ('Arbeitslosenhilfe') as well as social assistance ('Sozialhilfe') into a two-tier system of only unemployment benefits and social assistance ('ALG II'). Slightly deviating illustration from [Hochmuth et al. \(2021\)](#).

sents a drastic reduction in benefits. This is also reported by [Bradley & Kügler \(2019\)](#), who find that benefit payments were significantly reduced, especially for unskilled workers. This, however, was intended and the reform was designed to shift the focus from unemployment benefits as a form of insurance to incentives to take up work in such situations.

2.2 A simple bargaining model

This section draws on the previous narrative and theoretically relates the exogenous Hartz IV reform shock to labor market institutions such as unions. Drawing from the rent-sharing literature, we derive a model of union bargaining in which the threat point of unions is lowered because of the exogenous shock on alternative income stemming from the Hartz IV legislation. In this literature, there are direct and indirect factors affecting bargaining outcomes for workers in the labor market. Whereas direct factors increase the power of workers in negotiations (e.g. [Blanchard & Giavazzi 2003](#)), indi-

rect factors alter the outside option in case the negotiations break down. Our focus in this paper is on the latter.⁶

Explanations of the labor share based on bargaining power have to assume a rent-sharing framework in which economic rents at either the organizational or the country level have to be shared between capital and labor. Firms are therefore not price-takers and possess market power (e.g. De Loecker et al. 2020). In our model, however, markups do not arise due to market power in the product market, but rather through the power of unions which are able to shift the wage above its marginal product of labor.

We derive a simple model where we assume a market with two duopolists (firm 1 and firm 2) and a union which are involved in bargaining. The union's utility function is based on risk-neutral agents and is specified for the maximization of the rent of its members. The rent that employee's realize is the difference between the wage w and an alternative wage w_a . The monetary value of the alternative wage is determined either by unemployment benefits alone, or by a weighted average of (i) the wage when employed in another company and (ii) the unemployment benefit. The weights are the results of the employees' assessment of the probability of the two alternatives occurring. The value of w_a determines the lower limit of the negotiated wage w or the outside option in case negotiations fail. Thus, we focus in this paper on the so far overlooked relative change in workers' bargaining power. The introduction of the Hartz IV reform as outlined in Section 2.1, has led to a deterioration in financial support for large parts of the workforce. With the implementation of the Hartz IV reform, the alternative wage w_a therefore had decreased for parts of the workforce or at least is a very credible threat for lower wages when bargaining fails.

In the bargaining model, the rent per employee is multiplied by the number of employees who are members of the union. The aim of the union is to maximize the dif-

⁶There are in fact a few studies in this context which investigate indirect effects stemming from welfare services in several countries (e.g. Stockhammer 2017; Onaran 2009; Jayadev 2007). Although direct effects of bargaining power are not specifically considered in this paper, we include union density as a measurement of direct bargaining power in our regressions to adjust for this channel.

ference between the wage w and the alternative wage w_a which is simultaneously the threat point of the union. The term \bar{N} is considered as union membership in which not all employees from the pool of employment N have to be union members ($0 < N \leq \bar{N}$). We consider the following union's objective function of a utilitarian form:

$$U(w) = N(w - w_a) \quad (1)$$

The function in Equation (1) is the well-known Stone-Geary utility function with risk-neutral workers which is frequently applied in the literature (e.g. [Blanchflower et al. 1996](#); [Dobbelaere 2004](#)). We also assume that, in an event of bargaining delay, the firm earns zero profit because of the lack of workers, and employees receive the alternative wage w_a since they are unemployed for the time-being. Because of the dual structure of the industrial relations systems in Germany, unions at the industry-level usually bargain with employers' associations and determine wages but not employment. Thus, the bargaining framework in this paper considers wage bargaining, since the determination of employment levels is outside the scope of unions.⁷

Firm's utility is symmetric for firm i , where $i = 1, 2$ in this model and equals its profits π_i which is the output q_i times the price p . Output is produced using a Cobb-Douglas production technology in which we assume no fixed costs F and labor as the only variable input factor. Thus, the simple production function is $q_i = N$ in which the firm only has to pay the input costs w . The profit function then reads as follows:

$$\pi_i(w, N) = pq_i - wq_i \quad (2)$$

For pricing, the following linear inverse demand function is assumed:

$$p = d - b(q_1 + q_2) \quad (3)$$

⁷On the firm-level, however, wages are outside the field of application since they are determined on the industry-level by unions. In this level, firm owners or managers bargaining with co-determination institutions such as works council to determine employment. For a bargaining model on the company-level see for example [Kraft \(1998\)](#). For the empirical wage determination depending on different contracts between German unions and employer associations, see for example [Fitzenberger et al. \(2013\)](#).

As usual and shown in Equation (2), firms maximize the difference between sales and costs which leads to the following profit function for firm 1:

$$\begin{aligned}\pi_1 &= (d - b(q_1 + q_2))q_1 - wq_1 \\ &= (d - b(q_1 + q_2) - w)q_1\end{aligned}\quad (4)$$

We consider the more realistic case of asymmetric generalized bargaining power (e.g. Dobbelaere 2004; Blanchflower et al. 1996) in which the bargaining power of two players is denoted by ϕ for the union and $1 - \phi$ for the firm. The aims of the two parties are combined by the well-known Nash bargaining solution with Equation (1) and (4):

$$\Phi = (N(w - w_a))^\phi ((d - b(q_1 + q_2) - w)N)^{(1-\phi)} \quad (5)$$

For pure wage bargaining (and not efficient bargaining), N must be replaced by a function of w , respectively $N(w)$, before maximizing this Nash bargaining function. For this purpose, the profit function in Equation (4) is differentiated with respect to q_i and solved for output. Under the assumption of symmetric duopolists (with $q_1 = q_2$), this leads to:

$$q_1 = \frac{d - w}{3b} \quad (6)$$

This function is inserted into the bargaining Equation (5) and after taking the logarithm, this function reads:

$$\ln \Phi = \phi \ln \left(\frac{(w - w_a)(d - w)}{3b} \right) + (1 - \phi) \ln \left(\frac{(d - w)^2}{9b} \right) \quad (7)$$

From differentiation Equation (7) with respect to w (i.e. $\frac{\partial \ln \Phi}{\partial w}$) follows:

$$w = \frac{1}{2} (\phi(d - w_a)) + w_a \quad (8)$$

Unsurprisingly the negotiated wage w increases with w_a . Inserting Equation (8) for

w into the expression for q_1 which is Equation (6) gives the output and labor demand:

$$N = q_1 = \frac{(2 - \phi)(d - w_a)}{6b} \quad (9)$$

Output and thus also the demand for labor decrease with higher w_a . Then, in this Cournot model profits are given by:

$$\pi_1 = \frac{q_1^2}{b} = \frac{(2 - \phi)^2(d - w_a)^2}{36b} \quad (10)$$

Therefore as shown in Equation (10), profits also fall with w_a . We conclude from this simple model that in the case of a lowered outside option resulting from the Hartz IV legislation, wages will decrease, labor demand will increase and profits will rise. These results are consistent with existing empirical research (e.g. [Grüner 2019](#)).

The less obvious question, however, is what happens with the labor share if w_a falls. The labor share is defined as $ls = wN/pq$, and since in this simple model $q = N$, the expression reduces to $ls = w/p$. By use of Equations (3) and (9) the following expression for p can be derived:

$$p = d - \frac{1}{3}(2 - \phi)(d - w_a) \quad (11)$$

Then, after some rearrangement the labor share ls is now the following simple expression:

$$ls = \frac{\frac{3}{2}(\phi(d - w_a) + 2w_a)}{\phi(d - w_a) + 2w_a + d} \quad (12)$$

As a final step and to show the effect of changes in the alternative wage w_a , we take the derivative with respect to this coefficient:

$$\frac{\partial ls}{\partial w_a} = \frac{\frac{3}{2}(2 - \phi)d}{((2 - \phi)w_a + (1 + \phi)d)^2} > 0 \quad (13)$$

As evident from Equation (13) and outlined in the theoretical Section 2.1 in this paper,

the labor share ls falls with decreasing w_a .

3 The Hartz reforms as a structural break in the labor share

3.1 Data sources and measurement

We want to begin by identifying the Hartz IV reform, which was implemented on January 1st, 2005, as a significant break point in the time series of the German labor share. Tests for structural breaks in an economic context have a long history, starting with the early work by [Chow \(1960\)](#) and [Quandt \(1960\)](#). More recent theoretical contributions include [Bai & Perron \(1998\)](#); [Han & Park \(1989\)](#) as well as [Hansen \(2001\)](#), and nowadays there are many applications of change point tests in different fields (e.g. [Lunsford 2020](#); [Antoch et al. 2019](#); [Link & van Hasselt 2019](#); [Wiese 2014](#); [Jayachandran et al. 2010](#)). The general idea is to check whether an economic intervention constitutes a fundamental change in the data generating process and thus can be interpreted as a change point. We apply different exogenous and endogenous tests to check whether the Hartz IV reform in Germany in the year 2005 constitutes a significant impact on the labor share. Whereas in exogenous tests the year of the break point have to be explicitly defined, endogenous tests detect the break point from within the data.

We use the EU KLEMS⁸, revision 2019 dataset, which we combine with data from the Penn World Tables and the OECD STAN database. Then, we calculate the labor share LS_t for Germany for the years 1970 to 2015 as follows:

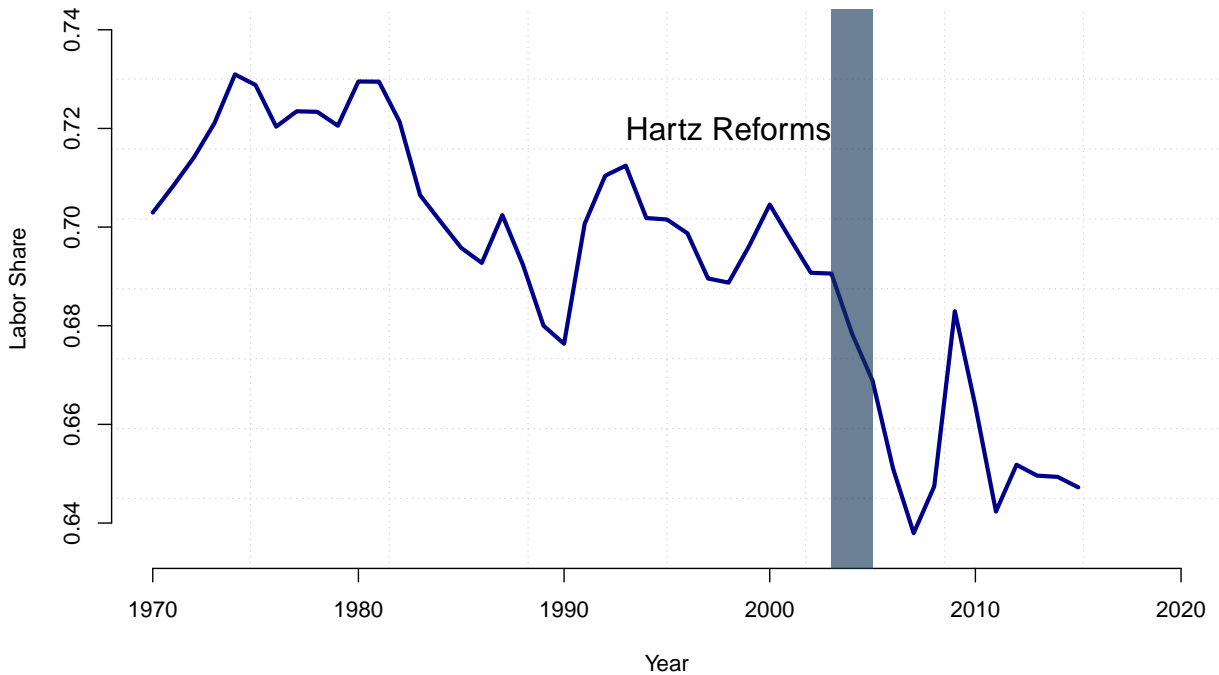
$$LS_t = \frac{W_t L_t}{Y_t} \quad (14)$$

where the expression $W_t L_t$ denotes labor compensation and Y_t gross value added at year t . One additional advantage of the EU KLEMS dataset is the consideration of self-employed workers who are often neglected when calculating the labor share (e.g.

⁸For a comprehensive discussion of the dataset and methodology, see [Stehrer et al. \(2019\)](#).

Cette et al. 2020). We therefore assume that the self-employed receive the same hourly wage as employees in every industry and year and thus prevent measurement errors in the labor share calculation (Stehrer et al. 2019).

Figure 2: Trend in the German labor share



Notes: This figure shows the trend in the German labor share LS_t over the years 1970 to 2015. The German labor share is calculated as the ratio of labor compensation to gross value added for all industries in year t as described in Equation (14). The data for the EU KLEMS release 2019 can be obtained on <https://euklems.eu/>. For an overview of variable construction and methodology see Stehrer et al. 2019. The blue shaded area indicates the implementation of the Hartz legislation in which Hartz I and II are implemented in 2003, Hartz III in 2004 and finally Hartz IV in 2005.

As shown in Figure 2, the aggregated labor share for Germany is in line with several other studies (e.g. Karabarbounis & Neiman 2014). There are three main events in recent decades which contribute to a change in the labor share in Germany. First, changes in the 70s can be attributed to the two oil price shocks in which the labor share sharply increases (Berthold et al. 2002). Second, the German reunification in 1990 also constitutes a sharp increase in the labor share because of large monetary transfers from western to eastern Germany in which the currency was not devalued.⁹ And, finally, the Hartz reforms contribute to a significant decline in the labor share in which the share stays rather constant hereafter. The only exception and sharp increase is due to

⁹For an synthetic control analysis of the effects of the German reunification on GDP, see for example Abadie et al. (2015).

the financial crisis in the year 2009, which can be explained by sticky wages and labor hoarding, which primarily affects capital incomes (Bazillier & Najman 2017). The literature suggests that many firms use working-time arrangements such as ‘time accounts’ or other work-sharing schemes (Teague & Roche 2014). Empirical studies indeed find that the labor share increases in times of economic downturn, as for example the great financial crisis in Germany of 2008–2009 (e.g. Bazillier & Najman 2017). Whether these changes, however, are significant in a statistical manner is the question which obviously presents itself. In the following we therefore test these hypotheses using structural break tests.

3.2 Supremum of a sequence of Wald tests

As it is apparent from Figure 2, there is a notable decline in the labor share around the Hartz legislation between the years 2003 and 2005. Around this period, Figure 2 suggests a change point in the mean of the labor share and of course, we could apply a simple t-test for pre- and post-reform mean differences using time dummy variables. However that would require that the break occurs at a known point in time. This is easy in principle, since we know the exact time of the implementation of the Hartz reforms and the potential break points.¹⁰ Choosing a fixed break date, however, might nevertheless be arbitrary since we do not know whether there are any delay or anticipation effects of the reform (e.g. Wiese 2014; Piehl et al. 2003).

As a more sophisticated approach, we apply endogenous tests for structural breaks in the mean for unknown break dates (e.g. Lunsford 2020; Wiese 2014; Jayachandran et al. 2010; Hansen 2001). The endogenous approach is much more reliable than the one with, for example, exogenously determined breakpoints because the endogenous test checks all possible alternatives. As a first test, we calculate Wald test statistics to determine whether there is indeed a structural break for a variety of break dates and take the maximum as the test statistic (Chow 1960; Quandt 1960). We test for a break in the mean of the labor share (LS_t) in year τ between $t = 1, \dots, T$ where $t = 1970$

¹⁰In such cases, the Chow test might be a feasible alternative (Chow 1960).

and $T = 2015$ estimating the following model several times¹¹ for every possible break point,

$$LS_t = \alpha + \delta_t D_t(\tau) + \gamma trend + \varepsilon_t \quad (15)$$

where $D_t(\tau)$ describes an indicator variable with $D_t(\tau) = 1$ if $t > \tau$ and $D_t(\tau) = 0$ otherwise. Thus, we test for all possible breaks in the mean for each year in the interval 1975 to 2009.¹² Given our stationary time series, which we confirm using a Dickey-Fuller test, we include a trend variable (e.g. [Rodriguez & Jayadev 2013](#)) and define the following Wald test statistic in which there is no change before and after the Hartz IV reform in the null hypothesis:

$$H_0 : \delta_t = \delta_0 \forall t, \\ H_1(\pi) : \delta_t = \begin{cases} \delta_1, & t = 1, \dots, T\pi \\ \delta_2, & t = T\pi + 1, \dots, T, \end{cases} \quad (16)$$

where the parameter $\pi \in (0, 1)$ is the sample fraction before and after the break point and $T\pi$ corresponds to the year of the change point. Then, we test the null hypothesis that there is no break point which is $\delta_t = 0$.

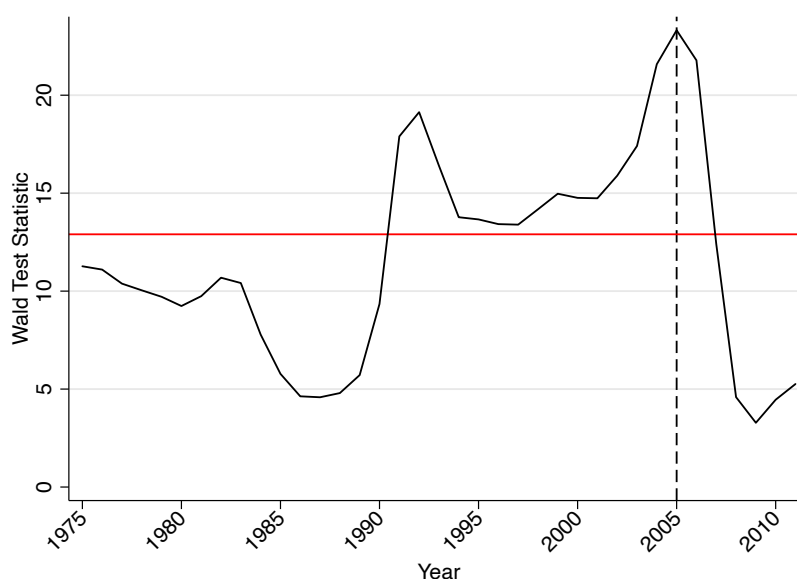
The maximum value of the Wald statistic (sup Wald) over all possible breaks is used to verify the break point and the significance of the break. [Figure 3](#) shows the values for every possible Wald test statistic for each year. The red line indicates the critical value provided by [Andrews \(1993, 2003\)](#) for the assessment of significance.

As shown in [Figure 3](#), the test statistic exceeds the critical value in the year of the German reunification in 1990. Hereafter, the test stays significant, indicating that the German reunification constitutes a rather lasting impact on the share of labor. The

¹¹We also estimated the model with successively added control variables in which we include the number of strike days, the unemployment rate as well as the growth rate of GDP. Results can be found in [Table A.4](#) and [Table A.5](#) of the Appendix.

¹²The test uses a slightly smaller sample size to ensure that the test has enough power. A common approach therefore is to trim 15 percent from both ends of the sample (e.g. [Jayachandran et al. 2010](#)).

Figure 3: Wald test statistic for structural break



Notes: The figure shows values of Wald test statistics as outlined in Equations (15) and (16). Test for a change point in the mean of the labor share (LS_t) in year τ between $t = 1, \dots, T$ where $t = 1975$ and $T = 2009$. The labor share is calculated as outlined in Equation (14). The blue line shows the values of the test statistics. The maximum value ($\sup \text{Wald}$) of the Wald test statistic is 23.32, which occurred in the year 2005. The red line indicates the critical value of the test statistics as provided by Andrews (1993, 2003). Values of the test statistics are provided in Table 1. We trim 15 percent from both ends of the sample to ensure that the test has sufficient power.

second break is indicated in the year 2005, which is also the maximum of the Wald test statistic ($\sup \text{Wald}$). It seems, therefore, that the Hartz IV reform is a profound structural break in the mean of the German labor share. Table 1 presents results for the applied single structural break tests. The p-values are calculated by the method provided by Hansen (1997) and the test statistic are derived and tabulated by Andrews (1993, 2003). In addition to the supremum Wald test we also apply an average Wald and exponential Wald test. These tests tend to have more power compared to the supremum test (Andrews & Ploberger 1994). Our results, however, do not change.

3.3 Bai and Perron test for multiple breaks

The in the previous section applied sequence of Wald tests are frequently used in the empirical literature (e.g. Lunsford 2020; Link & van Hasselt 2019; Jayachandran et al. 2010; Piehl et al. 2003; Hansen 2001), however these tests are limited to the occurrence

Table 1: Values for Wald test statistics

Test	Value of test statistic (1)	Break year (2)	p-value (3)
Supremum Wald	23.32	2005***	.0002
Average Wald	12.10		.0005
Exponential Wald	8.99		.0003

Notes: This table reports values of the Wald test statistics as outlined in Equations (15) and (16). Test for a change point in the mean of the labor share (LS_t) in year τ between $t = 1, \dots, T$ where $t = 1975$ and $T = 2009$. The labor share is calculated as outlined in Equation (14). Critical Values were obtained from Andrews (1993, 2003).

of only one break point. An inspection of the labor share time series in Figure 2 reveals, however, that there might in fact be more break points. We therefore expand the analysis and additionally allow for unknown timings and different numbers of change points in the labor share time series using the Bai & Perron (2003, 1998) test for multiple break points.¹³

The idea of the test is to create a step-by-step route through the adjusted labor share time series LS_t and create an optimal model with m breaks in $m + 1$ regimes. Drawing on a linear regression model (e.g. Casini & Perron 2019), we consider a model of the following form. In addition to a common method in which only a constant term is included (Wiese 2014), we also include the trend in the following regression model.

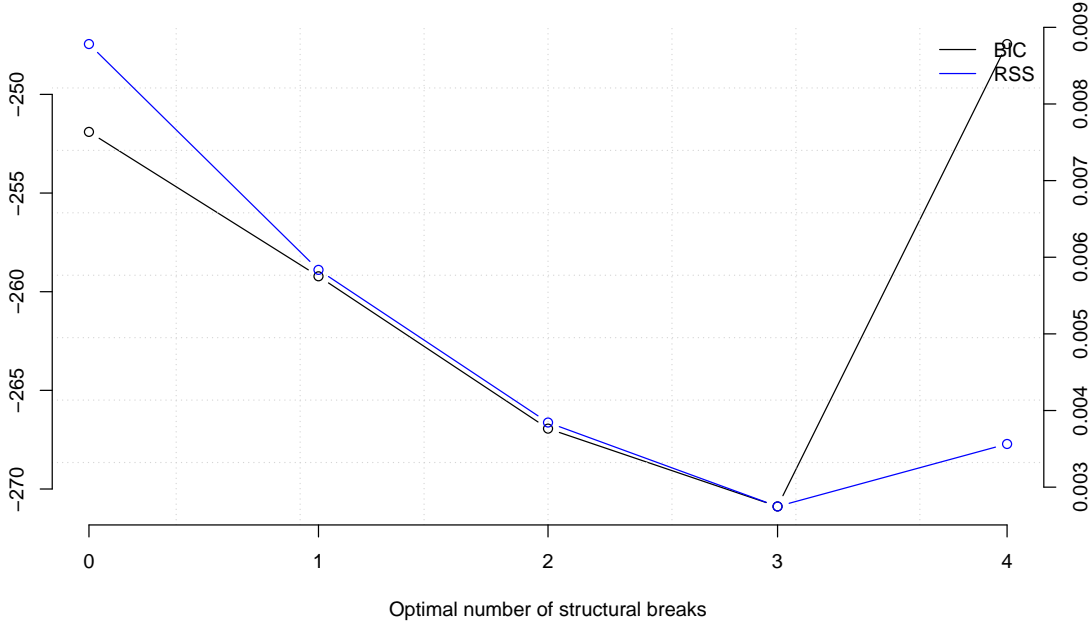
$$\begin{aligned}
 LS_t &= \delta_1 + \beta_1 x + u_t, & t = 1, 2, \dots, T_1 \\
 LS_t &= \delta_2 + \beta_2 x + u_t, & t = T_1 + 1, \dots, T_2 \\
 &\vdots \\
 LS_t &= \delta_m + \beta_m x + u_t, & t = T_m + 1, \dots, T
 \end{aligned} \tag{17}$$

in which the dependent variable is the labor share LS_t and δ_m being a vector of estimated constants of $m + 1$ possible regimes. Thus, it is the mean of the different segments which are divided into m breaks. The test then checks whether the change points are statistically significant. The number of break points is selected according to the low-

¹³See for example Wiese (2014); Benati (2007) for a similar application of the Bai & Perron test as well as Casini & Perron (2019) for a general assessment of structural break tests in time series.

est overall residual sum of square (RSS) for a given number of breaks and the Bayesian information criterion (BIC) which are shown in Figure 4. Both criteria refer to the optimal number of $m = 3$ breakpoints, dividing the labor share time series in Figure 2 into $m + 1 = 4$ regimes with different intercepts.

Figure 4: Optimal number of break points



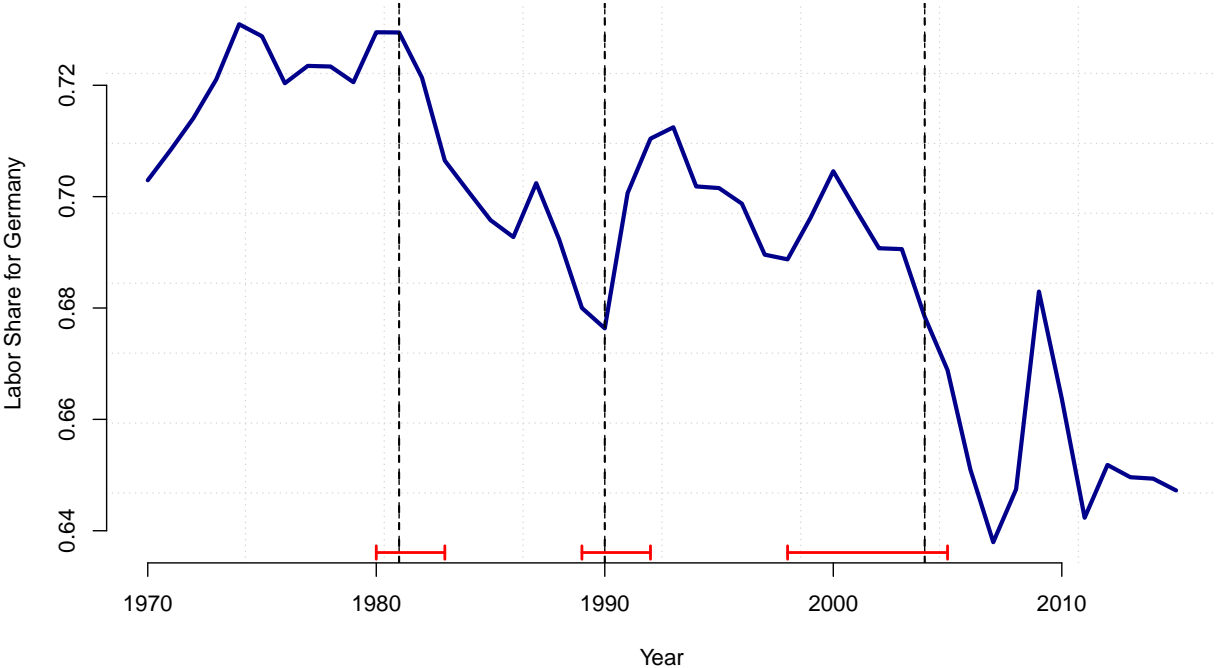
Notes: This figures shows the values of the selection criteria of the optimal number of unknown break points m in the (Bai & Perron 1998, 2003) test. According to the lowest residual sum of square (RSS) and Bayesian information criterion (BIC), the labor share times series in Figure 2 is divided into $m + 1 = 4$ regimes as shown in Figure 5.

The Bai & Perron (1998, 2003) test has different sequential stages, which will be briefly outlined. First, a supF type test is carried out to determine whether there is no structural break ($m = 0$) at all or a fixed number of breaks ($m = k$). In the next step, the null hypothesis that no structural break is present is tested versus the alternative hypothesis of an unknown number of structural breaks, with an upper limit being set. This test is implemented with double maximum tests. The first of these tests is based on equal weighting, while the second uses weights for the individual tests, which are calculated in a way that the marginal p-values are equal across values of m . The weighting is implemented because with an equal weighting the power of the test decreases when the number of structural breaks increases.

The next step is to identify the optimal number of structural breaks. Bai & Perron

(1998, 2003) propose a test for a particular number of structural breaks l versus $l + 1$. The corresponding test $\sup F_t(l+1|l)$ then gives the maximum of the F-statistic, which tests the null hypothesis that no additional structural break exists versus the existence of an additional structural break. This test is performed sequentially for all possible points in time. The optimal number of structural breaks is then identified using residual sum of square (RSS) and the Bayesian information criterion (BIC). The approach allows for non-symmetric confidence intervals since the variance before and after a break does not have to be constant. Furthermore, the variance covariance matrix is robust to serial correlation and heteroscedasticity (Andrews 1991). The results of the described sequential testing procedure are summarized in Figure 4. The number of breaks is plotted on the horizontal axis, while the two vertical axes show the values of the BIC statistics and the residual sum of squares. The BIC has its minimum at three breaks.

Figure 5: German labor share with identified break points



Notes: Own calculations using the EU KLEMS 2019 release. The German labor share over the year 1970–2015 is calculated as the ratio of labor compensation to gross value added for all industries as shown in Equation (14). Dotted lines show endogenously identified change points using the Bai & Perron (1998, 2003) test explained above. Red bars at the bottom indicate 95 % confidence intervals around the estimated break point. The break points occur after the oil crisis in 1981, in 1990 during the period of East and West German reunification as well as in the year 2004 during the implementation of the Hartz reforms.

Figure 5 shows the exact times identified by the Bai & Perron (1998, 2003) test. We first find that these results are consistent with our visual assessment based on the previous graphs. For two breaks, the Bai-Perron estimates identify the same time points that the previous statistics found. Second, further change points that we found in this time series are also in line with theoretical predictions from the literature. The first break occurs right after the second oil price crisis in the year 1981 in which the oil prices rapidly fell (Autor et al. 2020). Thus, the severe drop in oil prices might spur capital intensive production thus leading to a decline in the labor share. The second break occurs shortly after the East and West German reunification in 1990 and the third break occurs at the time of the introduction of the Hartz reforms. The 95 % confidence intervals are shown in Figure 5 around the break points. Because of our rather short time series, since we use yearly data compared to daily or monthly data, for example, the confidence intervals are quite broad. Nevertheless, the breakpoints are consistent with the previous analysis of Wald statistics, which is summarized in Table 1. Important to note in our analysis, however, is the fact that these change points are endogenously identified within the labor share time series and interestingly they also highlight the Hartz reforms as having a significant impact.

To summarize, we use an endogenous Wald test for a single break point as well as the Bai & Perron (2003) test for the identification of multiple endogenous break points. We indeed verify that the Hartz reforms in Germany constitute a significant shift in the mean of the labor share. This result can be explained by the fact that the introduction of the Hartz IV reform has reduced the outside option in our bargaining framework for employees and therefore had a negative impact on the outcome in wage negotiations. The macroeconomic analysis also has the advantage that tests on structural breaks can be carried out without predetermined break points and with several breaks.

4 Impact of the Hartz IV reform

4.1 First results using aggregated data

The structural break tests indeed indicate a significant break right before the Hartz IV legislation in Germany became effective in the year 2005. In this section we therefore inspect the effect of this exogenous shock of the Hartz IV reform on the aggregated labor share. As outlined in Section 2.2, Equation (13), we expect that the exogenous negative shock on the outside option in the bargaining equation subsequently results in lower wages and, therefore, in a decrease in the labor share. To examine our hypothesis, we first apply ordinary least squares (OLS) regressions of the following form:¹⁴

$$LS_t = \alpha + \beta_1 HartzIV_t + \beta_2 Unemp + \beta_3 Strike + \beta_4 Union + \beta_5 ExImRatio + \beta_6 trend + \varepsilon_t \quad (18)$$

Where LS_t is the labor share in Germany ranging from $t = 1970$ to $T = 2013$. As control variables we include the trade union density (*Union*), which is a measure for the bargaining power of employees.¹⁵ In fact, there is much literature about the impact of labor unions on the distribution of incomes and factor shares by different channels. First, there are direct positive effects stemming from bargaining in which labor unions reduce within-group as well as between-group wage inequality (e.g. [Kristal & Cohen 2017](#)). Furthermore, and more in line with our research, a strand of literature suggests that labor unions affect the compensation of the management and also returns to capital (e.g. [Lee & Mas 2012](#)). Second, a more recent paper suggests that labor unions do not affect the wages of employees directly, but rather that positive distributional effects arise from more generous fringe benefits (e.g. [Knepper 2020](#)). In this context, [Card et al. \(2020\)](#) provide a very recent overview on labor unions and inequality.

In addition, we also include the unemployment rate of Germany (*Unemp*). A higher

¹⁴Results from regressing the labor share on year dummy variables is provided in Figure A.6.

¹⁵Trade union density is measured as the members in the German federation of trade unions in the corresponding year (which can be found here: <https://www.dgb.de/uber-uns/dgb-heute/mitgliederzahlen>) over the total employment as reported in the EU KLEMS dataset.

unemployment rate represents an increased risk for workers of having to search longer for new employment in the event of a job loss, and therefore of being at least temporarily affected by unemployment themselves. This in turn will reduce bargaining power. We also include lost workdays due to strike in 1,000 employees, *strike* as an additional measure for workers' bargaining power. Finally, the export-import ratio (*ExImRatio*) is included, which is a measure for trade openness and globalization (e.g. [Elsby et al. 2013](#)). We expect this variable to carry a negative sign since there is much evidence that the relationship between globalization and the labor share is negative ([Elsby et al. 2013](#)).

Table 2 shows the descriptive statistics for the EU KLEMS data which is used for estimating Equation (18).

Table 2: Descriptive statistics for German EU KLEMS sample (1970–2013)

Variables	Mean (1)	Std. dev. (2)	Min (3)	Max (4)
Labor share	0.69	0.03	0.64	0.73
Unemployment rate	7.09	3.12	0.58	11.72
Trade union density	0.32	0.04	0.24	0.41
Export-import ratio	1.00	0.11	0.80	1.19
Lost workdays	24.05	60.01	0.40	278.60
Year			1970	2013
Observations	44			

Notes: The table shows the descriptive statistics for the German labor share sample ranging from 1970 to 2013. Data from the EU KLEMS release 2019 dataset and trade union information from the German federation of trade unions. Lost workdays due to strike from the Hans Böckler Foundation, per 1,000 employees. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. For an overview of variable construction and methodology see [Stehrer et al. \(2019\)](#).

First results from estimating Equation (18) using OLS are presented in Table 3 with additionally adding control variables in column (2). As expected and derived in the theoretical model, the Hartz IV reform has a significantly negative impact on the labor share in the magnitude of -2.7 percentage points when adjusting for control variables. Regarding the control variable as shown in column (2) of Table 3, the unemployment rate also carries a negative sign as described above. The results are also similar when

Table 3: Results for OLS regression for the German labor share

	Labor Share	
	(OLS) (1)	(OLS) (2)
Hartz IV	-.028*** (.006)	-.016* (.009)
Unemployment rate		-.001 (.001)
Lost workdays		-.016 (.018)
Trade union density		.191** (.090)
Export-import ratio		-.061 (.037)
Linear trend	-.001*** (.000)	-.001* (.000)
Constant	.725*** (.004)	.728*** (.043)
R^2	.817	.838
Observations	44	44

Notes: This table show results from estimating Equation (18). Data from the EU KLEMS release 2019 dataset and trade union information from the German DGB trade union association. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. For an overview of variable construction and methodology see Stehrer et al. (2019). Results from regressing the labor share on year dummies is provided in Figure A.6. Robust standard errors in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level.

we use year dummy variables instead of a Hartz IV indicator variable as shown in Figure A.6 of the Appendix. The coefficient on trade union density has an expected positive sign and measures the impact of an increase in bargaining power of employees on the labor share (e.g. Brock & Dobbelaere 2006). Finally, the measure of globalization (export-import ratio) also has a negative coefficient. As expected and shown in various studies, trade openness and globalization lead to more offshoring of labor-intensive work and thus a decrease in the labor share (e.g. Elsbey et al. 2013).

4.2 Synthetic control method using a counterfactual Germany

By application of endogenous structural break models and simple OLS regressions as shown in Table 3, our results presented so far support the hypothesis that the Hartz reforms affected the labor share. In the next step we take possible selectivity effects into account.

Considering causal models, there is the well-known fundamental problem of causal inference which states that it is only possible to observe outcomes for entities (such as countries or firms) which are either treated or untreated (e.g. Rubin 1974; Imbens & Wooldridge 2009). A clearly defined control group with similar characteristics as the treatment group is therefore needed to draw causal conclusions. In our specific case, however, the Hartz reforms affected the whole economy in Germany and the labor share, and there is no natural control group. We therefore apply a more advanced form of structural break analysis by using the synthetic control method (e.g. Abadie et al. 2010, 2015; Abadie 2021). This method is not only more in line with recent developments in econometrics (e.g. Imbens & Wooldridge 2009) but also referred to as one of the most important innovation in the recent policy evaluation literature (Athey & Imbens 2017). In fact, there is an ongoing development of this method regarding the implementation of covariates (e.g. Botosaru & Ferman 2019) as well as many empirical applications in different fields of economics (Chen 2020; Peri & Yasenov 2019)

Data and sample construction. For the application of the synthetic control method and the construction of a ‘synthetic twin of Germany’ the analysis requires additional data on other countries. These countries are referred to as the donor pool to construct the German doppelganger. We therefore combine the EU KLEMS dataset with the Penn World Tables 9.0 database¹⁶ for information regarding GDP spending and the OECD STAN database for trade union density. Unfortunately, we do not have such rich information for other countries that allow us to create such a long time series as

¹⁶The Penn World Table database 9.0 covers information on relative levels of income, output, input and productivity for 182 countries between 1950 and 2014. For an overview of the methodology as well as variables see for example Feenstra et al. (2015). Missing data on the unemployment rate for various countries were supplemented by World Bank data.

we have for Germany starting in the 1970s. Our analysis in this section therefore starts in the year 1995, for which we have all information on control variables.

According to the literature, we use quite common variables to construct the Germany doppelgänger. First, trade union density and the unemployment rate are included as a measure of bargaining power (Blanchard & Giavazzi 2003; Bental & Demougin 2010). Also, the share of capital formation (as % of GDP) is included as a measure of capital accumulation in the economy which is relevant to account for production capacities (Piketty & Zucman 2014). The share of government consumption as a proxy for the welfare state (i.e. a proxy for social protection) which is, for example, also applied by Bazillier & Najman (2017). As a measure for globalization, we include the share of exports and imports (as % of GDP) (Elsby et al. 2013). And, finally, we account for the stock and quality of human capital in the economy by using the ‘human capital index’ provided by the Penn World Tables dataset.¹⁷ Descriptive statistics are provided in Table 4.

Methodology. Following the notation of Abadie et al. (2010), we apply the synthetic control method for the case that a single unit (Germany) is exposed to some treatment (the Hartz IV legislation) and there is no natural control group. The other ($J + 1$) countries remain unexposed to the reform and are referred to as the donor pool. This pool is then used to construct the Germany doppelgänger as a counterfactual Germany without the Hartz legislation. Basically, the idea of this approach is to build a counterfactual Germany without the Hartz IV reform from the donor pool of the other $N = 22$ countries¹⁸.

We observe the labor share for T periods in which the intervention starts in some period $T_0 + 1$ which is in our case the year 2005. Furthermore, we define the outcomes

¹⁷For example, firms need human capital to innovate and improve existing technologies which, in turn, affects capital and the production process. Thus, recent lines of research suggest accounting for this measurement in labor share regressions (e.g. Arif 2021).

¹⁸The countries that comprise the donor pool are in alphabetical order: Austria, Belgium, Czech Republic, Denmark, Estonia, Finland, France, Germany, Greece, Hungary, Ireland, Italy, Japan, Latvia, Lithuania, Luxembourg, Netherlands, Poland, Slovenia, Spain, Sweden, United Kingdom and the United States.

Table 4: Descriptive statistics for EU / World sample

Variables	Mean (1)	Std. dev. (2)	Min (3)	Max (4)
Labor share	0.62	0.06	0.49	0.86
Trade union density	32.35	21.20	5.50	95.80
Unemployment rate	8.27	4.05	2.25	27.48
Human capital index	3.21	0.27	2.55	3.73
Share of gross capital formation	0.25	0.05	0.10	0.47
Share of government consumption	0.19	0.05	0.09	0.42
Share of exports	0.48	0.27	0.08	1.39
Share of import	-0.51	0.28	-1.47	-0.11
Year			1995	2015
Observations	412			

Notes: The table shows the descriptive statistics for the shorter but more comprehensive country panel dataset. Data from the EU KLEMS release 2019, Penn World Table 9.0, OECD STAN as well as World Bank. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3 for each country. Figure A.1 on the Appendix provides the labor share trends for each country. Data sets are merged using country names and the corresponding year. For an overview of the EU KLEMS data see Stehrer et al. (2019) and for Penn World tables see Feenstra et al. (2015). More information regarding the human capital index is provided in Table A.1. The sample comprises the countries: Austria, Belgium, Czech Republic, Denmark, Estonia, Finland, France, Greece, Hungary, Ireland, Italy, Japan, Latvia, Lithuania, Luxembourg, Netherlands, Poland, Slovenia, Spain, Sweden, United Kingdom and the United States.

Y_{jt} which are the observed outcomes for the treated as well as the control countries. The fundamental problem in such a causal analysis is that we do not observe the labor share Y_{1t} for a counterfactual Germany, without the Hartz IV reform, after period T_0 . Fortunately, the synthetic control method offers a solution for the estimation of Y_{1t} , by creating a ‘synthetic control Germany’ as a weighted combination of other countries ($w_j Y_{jt}$) which best approximate relevant pre-intervention variables. The weighting vector is defined as $W = (w_2, \dots, w_{(J+1)})$ in which w_j is the contribution of each of the $N = 22$ donor pool countries.¹⁹ The counterfactual Germany is constructed as a convex combination of the observed outcomes of the other countries and there should be no difference in the real and the synthetic Germany prior to the intervention:

$$Y_{1t} = \sum_{j=2}^{J+1} w_j Y_{jt}, t = 1, \dots, T_0 \quad (19)$$

¹⁹Since Germany is the treatment state it is not part of the weighting vector. Moreover, the weights are constrained in that $w_j \geq 0$ and they sum up to one ($w_2 + \dots + w_{(J+1)} = 1$).

The effect of the intervention for the aggregated time period after T_0 is then obtained by the difference of Equation (19) for the time period after the intervention compared to prior to the intervention which is $\delta_{jt} = Y_{1t} - \sum_{j=2}^{J+1} w_j Y_{jt}$. This can be estimated using a difference-in-differences (DiD) framework in which we regress the difference between the real and the synthetic Germany on a Hartz IV variable which takes unit value for all years after 2004 and zero otherwise. This synthetic control method derives the well-known average treatment effect on the treated (e.g. [Abadie 2021](#)).

Country weights and counterfactual Germany. As shown in Table 5 there are four states which in particular resemble the German trend in the labor share quite well. These countries receive the corresponding weight which is shown in Table 5 for the construction of the counterfactual Germany as shown in Equation (19). Thus, using the synthetic control method with covariates (e.g. [Botosaru & Ferman 2019](#)) as provided in Table 4 for the construction of the Germany doppelganger, the Netherlands, Spain, Slovenia as well as the United Kingdom in particular are used as weighted aggregated comparisons.

Balancing of covariates. Similar to other matching algorithms such as propensity score matching (e.g. [Caliendo & Kopeinig 2008](#)), the credibility of the synthetic control method also relies on covariate balancing. The balancing of the pre-intervention variable means should be checked, which is done in Table 6 by comparing the means between the weighted control group and treated Germany. In our case, we construct the synthetic Germany as a convex combination of the 22 donor pool states that resemble Germany as closely as possible in terms of pre-intervention variables. As becomes apparent in Table 6, the means between real and synthetic Germany are balanced.

Results of the synthetic control method. Results from the synthetic control approach in addition to the trends among the synthetic and real Germany, are presented in Figure 6. Depicted in Table 6, the pre-intervention differences are balanced and trends between both labor shares are quite similar prior to the Hartz IV treatment. The Hartz

Table 5: Country weights from donor pool

Country	Weight	Country	Weight
Austria	< 0.01	Japan	< 0.01
Belgium	< 0.01	Latvia	< 0.01
Czech Republic	< 0.01	Lithuania	< 0.01
Denmark	< 0.01	Luxembourg	< 0.01
Estonia	< 0.01	Netherlands	0.276
Finland	< 0.01	Poland	< 0.01
France	< 0.01	Spain	0.273
Greece	< 0.01	Slovenia	0.226
Hungary	< 0.01	Sweden	< 0.01
Ireland	< 0.01	United Kingdom	0.225
Italy	< 0.01	United States	< 0.01

Notes: This table shows how the weighting vector in Equation (19) is resembled and which weights (i.e. $W = (w_2, \dots, w_{(J+1)})$) each donor pool country receives for the construction of the counterfactual Germany. Within the synthetic control method, variables are averaged for the pre-intervention period 1995 – 2004 (the share of capital formation, government spending, export as well as imports are averaged 1995 – 1999). Own calculations using the parametric synthetic control approach (e.g. Abadie et al. 2010, 2015; Abadie 2021).

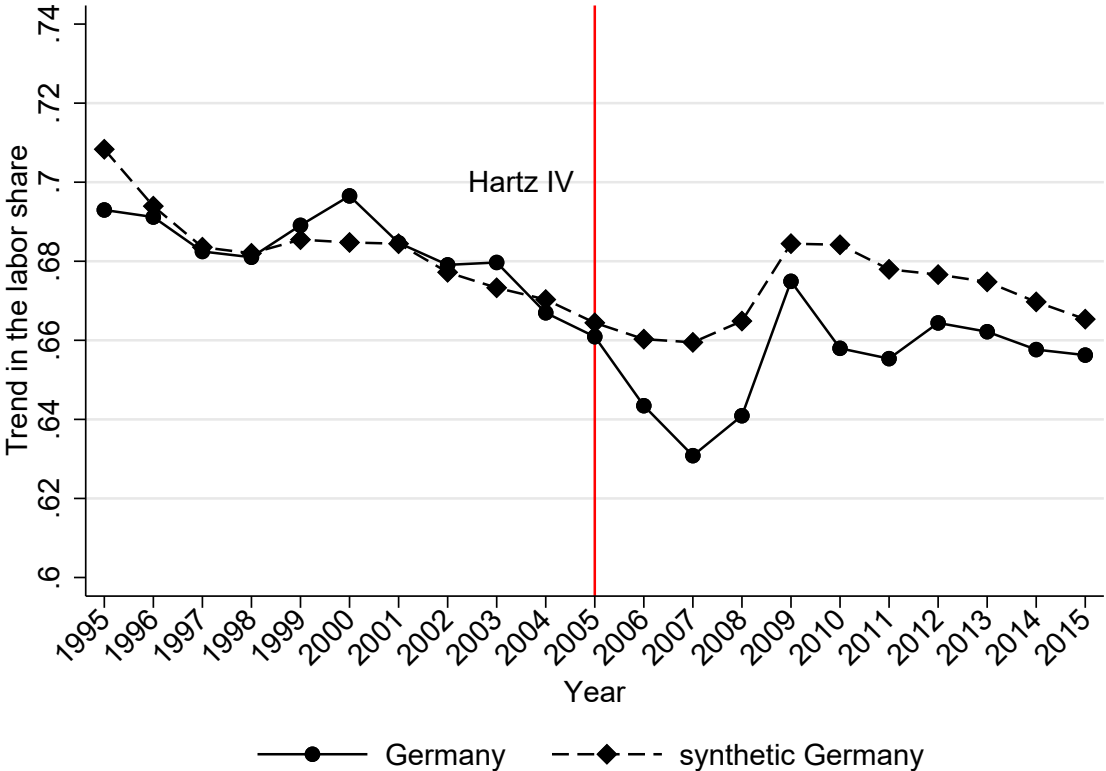
Table 6: Labor share predictor means before intervention

Covariates	Germany	
	Real (1)	Synthetic (2)
Labor share	.684	.684
Trade union density	25.22	28.06
Unemployment rate	8.81	8.44
Human capital index	3.55	3.10
Share of capital formation	.251	.252
Share of government consumption	.137	.160
Share of exports	.356	.356
Share of imports	-.355	-.390

Notes: This table shows mean comparisons of covariates used for the calculation of the counterfactual labor share which is shown in Figure 6 and the real German labor share. Means for the synthetic Germany are calculated using the weights provided in Table 5. For the construction of the weights, all variables are averaged for the pre-intervention period 1995–2004 (the share of capital formation, government spending, export as well as imports are averaged 1995–1999).

IV reform was enacted at the beginning of the year 2005 in which we see a sharp decline in the German labor share; however, not in the counterfactual trend. Interestingly, both trends capture the effects of the financial crisis in the year 2009 very well, however the real German increase is much more pronounced. What is additionally important besides the strong decline in the real Germany, is the fact that the differences among the labor share are persistent after the Hartz IV legislation.

Figure 6: Trends in labor shares: real vs. synthetic Germany



Notes: The figure shows the trends in the real and the synthetic labor share of Germany. The labor share is calculated as outlined in Equation (14). All variables are averaged for the pre-intervention period 1995–2004 period (the share of capital formation, government spending, export as well as imports are averaged 1995–1999). Data from the EU KLEMS, Penn World Tables, OECD STAN as well as World bank.

Using the synthetic control approach, we are able to compare the real Germany where the Hartz IV legislation was enacted, with our synthetic control Germany doppeleganger which never experienced the Hartz IV reform. Our graphical results in Figure 6 show the significant negative impact of the Hartz IV reforms for the aggregated real labor share in Germany compared to the synthetic labor share. Regression results

using a Hartz IV dummy variable which takes unit value after the year 2005 and zero otherwise, yield similar results as before. The negative reform effect is significant and in the magnitude of 2.7 percentage points. Compared to the estimates using the longer time series for Germany only, as presented in Table 3, the effect of the synthetic method is slightly smaller, but also highly significant. The decline in the labor share is roughly 1.6 percentage points after the Hartz IV reform was implemented at the beginning of 2005. The comparison of Germany with the synthetic control group shows that both follow the same downward trend in the labor share over some time, but in contrast to the control observation, this trend became stronger for Germany after the Hartz reforms. Moreover, the impact of the reform is strongest in the early years before the great financial crisis in the year 2009, although the difference remains, albeit slightly smaller in terms of magnitude.

While the estimates provided in this section support our theoretical model and are also supported by the change point tests, there is still a lack of a cross-sectional dimension. It is quite likely that industries and firms are affected differently, in particular when estimates are additionally adjusted by control variables. For example, the decline in the degree of trade union organization or the increasing decentralization of the negotiation process (e.g. [Dustmann et al. 2014](#)) may explain the low wage increases, and in fact the often discussed real wage reductions in the 2000s. Since variables vary between sectors and firms, we carry out additional research using OLS, fixed effects as well as system GMM estimation to provide comprehensive firm-level evidence.

5 Robustness using firm-level evidence

5.1 Benchmark results.

In this section we provide evidence on labor share changes stemming from the firm level (e.g. [Harju et al. 2021](#)). The use of firm-level data has two main advantages. First, we are able to take changing sector and industry compositions into account that may affect the labor share ([Siegenthaler & Stucki 2015](#)). Second, as mentioned by [Gollin](#)

(2002) as well as [Karabarbounis & Neiman \(2014\)](#), aggregate labor share measures may be confounded by capital incomes earned by entrepreneurs and sole proprietors. We use firm-level data stemming from the ‘dafne’ dataset compiled by Bureau van Dijk, which allows us to measure the impact of a change in the outside option more precisely. We use information for the years 2000–2011.

Dependent variable. For the dependent variable, we use the same variables as explained earlier in Equation (14). Thus, we measure the labor share as the share of labor compensation to value added of the firm. We measure value added as the gross output minus intermediate inputs, depreciation as well as interest expenses. See Figure A.7 for a histogram of the labor share and Figure A.8 for the development of the firm-level labor share over time. Comparing the firm-level labor share with the aggregated German labor share, we see a quite similar development; a sharp decline around the Hartz legislation as well as a spike during the financial crisis.

Control variables. With respect to control variables, we implement variables which are common in the literature as determinants of the labor share. First, in neoclassical growth models, the labor share is a function of the capital-output ratio ([Bentolila & Saint-Paul 2003](#); [Siegenthaler & Stucki 2015](#)). Thus, we implement the logarithm of the capital-to-output ratio as an explanatory variable. The sign of the capital-to-output ratio depends on the elasticity of substitutions between capital and labor ([Siegenthaler & Stucki 2015](#)). In the case where labor and capital are complements, the capital-output-ratio increases the labor share. In the case of a substitutive relationship between capital and labor, the capital-output ratio decreases the labor share. The empirical literature usually finds controversial signs in different studies.

To account for unobserved demand shocks within the regression framework, we also add the logarithm of the ratio of intermediate inputs to firms’ value added and its square to the regression equation. The idea is to take changes in the input factors into account when firms are hit by various demand shocks (e.g. [Levinsohn & Petrin 2003](#); [Siegenthaler & Stucki 2015](#)). The labor share also depends on the degree of organiza-

tion of the workforce which we measure using union density. Given the relevance of collective bargaining agreements as well as the trend of declining unionization (Oberfichtner & Schnabel 2019) we expect lower wages (e.g. Akyol et al. 2013). In the case of a rather inelastic demand curve this also implies a reduction of the labor share (Siegenthaler & Stucki 2015). With respect to the sign of the coefficient, we therefore expect a positive sign of our union measure. With respect to the union measure we calculated a unique measure for union density at the industry level. See Section A.3 in the Appendix for the construction of the index. We also include detailed NACE (rev. 2.0) two-digit industry-fixed effects as well as year-fixed effects in the regression model. Finally, we include further control variables subsumed in X_{it} . The vector includes an indicator variable whether the firm is a stock company or a Societas Europaea (SE) and whether the firm is located in Western Germany. Descriptive statistics are provided in Table 7.

Table 7: Descriptive statistics for firm-level data

Variables	Mean (1)	Std. dev. (2)	Min (3)	Max (4)
Laborshare	.456	.173	.001	.998
ln (Capital-output ratio)	.139	.271	-5.22	4.17
ln (Intermediate inputs)	.689	.546	-.559	6.95
ln (Intermediate inputs squared)	.769	1.55	.001	48.42
Trade union density	.194	.095	.008	.960
No stock company	.736	.440	0	1
Stock company	.261	.439	0	1
Societas Europaea (SE)	.002	.048	0	1
Western Germany	.762	.425	0	1
Year			2000	2011
Observations	38,808			

Notes: The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. Trade union density is measured at the industry level as outlined in Section A.3.

Regression framework. We apply the following regression framework by using the logarithm of the labor share $\ln(LS_{it})$ as the dependent variable for firm i in year t as

follows:

$$\begin{aligned} \ln(LS_{it}) = & \beta_0 + \beta_1 \ln(k_{it}) + \beta_2 \text{Unemploy}_j^{2002} + \tau \text{HartzIV} \times \text{Unemploy}_j^{2002} \\ & + \beta_3 \text{Org}_{it} + \beta_4 \mathbf{Z}_{it} + \beta_5 \mathbf{X}_{it} + \mu_i + \theta_t + \varepsilon_{it} \end{aligned} \quad (20)$$

In this specification the labor share depends on the capital-output ratio (k_{it}), the vector \mathbf{Z}_{it} subsumed intermediate inputs of the firm as well as its square, Org_{it} measure bargaining power as union density and we include a set of firm specific control variables (\mathbf{X}_{it}). Finally, we insert time dummies θ_t to capture year specific shocks as well as NACE 2.0 industry classification dummy variables μ_i . The German unemployment rate is measured on the county-level j as shown in Figure A.9 in the Appendix. We estimate this specification using ordinary least squares (OLS), random effects (RE) as well as fixed effects (FE) models shown in Table 8. Results without the unemployment interaction term are quite similar and provided in Table A.2 in the Appendix as well.

As a final robustness check and to take potential endogeneity into account, we estimate System-GMM models (Blundell & Bond 1998). Therefore, we estimate dynamic models of the labor share with lagged levels of the labor share in the dynamic setting.²⁰ Regarding the generated instruments within the System GMM approach, we stack them into one vector to prevent the problem of weak instruments. To test the validity of our instrumental variables, we apply the standard Hansen test to check over-identifying restrictions (Hansen 1982). In addition, the system GMM estimator requires the absence of second-order auto-correlation in the residuals. We therefore also test for first- AR(1) and second-order AR(2) auto-correlation (Arellano & Bond 1991). With respect to standard errors, we use the standard two-step clustering approach and apply the Windmeijer (2005) finite sample correction.

Identification. Similar as described in the previous section, the Hartz reforms were introduced simultaneously, which implies the lack of a natural control group. While the previous section deals with this obstacle using a synthetic control approach, the

²⁰For a similar approach, see Böckerman & Maliranta (2012); Yang & Tsou (2021).

identification on the firm level relies on county-level variation in the unemployment rate before the Hartz reforms. Therefore, a variant of a difference-in-differences framework is applied (e.g. Card 1992; Giebel & Kraft 2019). In particular variation in the unemployment rate in the year 2002 is used as an exogenous measure for treatment intensity before the Hartz reforms were introduced. Thus, the effect of the Hartz reform is expected to be stronger in case the value of this variable, i.e. the unemployment rate increases. Basically, a higher unemployment rate implies less attractive outside options which lowers the labor share at an increasing rate. For the distribution of the unemployment rate by German counties, see Figure A.9 in the Appendix. Similar to the common difference-in-differences model, the treatment variable, i.e. the Hartz IV dummy variable, is included in the regression unchanged as before and interacted with the county-level unemployment rate in the year 2002.²¹ Thus, the $HartzIV \times Unemploy_j^{2002}$ coefficient takes the value of the corresponding unemployment rate of the county the firm is located in and zero for all years before 2004.

Results. Similar to our initial results for Germany, as well as the more sophisticated synthetic control approach, we also find a reduction of the labor share on the firm level stemming from the Hartz IV legislation. The size of reduction is similar to the magnitude found in the regressions on the aggregated labor share. These firm-level results are consistent among pooled OLS, random and fixed effects as well as system GMM estimation.²² With respect to the latter estimation results, the test statistics also support our results. After adjusting for time specific shocks in the labor share series using year-fixed effects, the Hartz IV dummy variable points to a significant negative relationship. Thus, an increase in the unemployment rate by 1 % results in a reduction of the labor share by 0.62 percentage points (OLS) to 0.228 percentage points (System-GMM). The point estimates are consistent among specifications and similar to the findings on the aggregated labor share. In line with the literature (e.g. Arif 2021; Siegenthaler &

²¹Since we use firm-level data, firms are matched with the county-level unemployment rate depending on the postal code using NUTS 3 classification to match the postal code with the corresponding county identifier.

²²We also applied different specifications in which the results are consistent among specifications.

Table 8: Results with treatment intensity on the firm level

	OLS	RE	FE	Sys-GMM
Lag labor share				0.647*** (0.092)
Unemploy	0.001 (0.001)	0.001 (0.001)		0.002** (0.001)
Hartz IV \times Unemploy	-0.00062** (0.00029)	-0.00083*** (0.00020)	-0.00086*** (0.00021)	-0.00228** (0.00100)
Capital-output ratio	0.039*** (0.006)	0.066*** (0.007)	0.076*** (0.009)	-0.033 (0.081)
Intermediate inputs	0.009 (0.006)	0.069*** (0.009)	0.102*** (0.011)	0.330** (0.131)
Trade union density	0.010** (0.005)	0.008** (0.003)	0.007** (0.003)	-0.019 (0.028)
Intermediate inputs squared	-0.011*** (0.002)	-0.008** (0.004)	-0.006 (0.004)	-0.116** (0.050)
Western Germany	0.019*** (0.006)	0.021*** (0.006)	0.000	0.009** (0.005)
Constant	0.335*** (0.018)	0.252*** (0.016)	0.319*** (0.011)	-0.163 (0.107)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
Limited liability controls	✓	✓	✓	✓
R2	.164	.130	.137	
No. of instruments				115
AR1 (p-value)				.000
AR2 (p-value)				.787
Hansen-J (p-value)				.942
Observations.	36,789	36,789	36,789	27,934

Notes: This table shows results from estimating Equation (20). The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) auto-correlation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

Stucki 2015), we also find that labor and capital in our data are complements (measured by the positive impact of capital-output ratio). With respect to our measure of trade union density, we find a positive association between bargaining power and the labor share as, for example, also found by Stansbury & Summers (2020). Regarding the measurement of intermediate inputs, our results are also in line with the findings by Siegenthaler & Stucki (2015) in which we also see an inverted u-shaped relationship.

Table 9: Results for laborshare firm level

	OLS	RE	FE	Sys-GMM
Lag labor share				0.651*** (0.083)
Hartz IV \times East	-0.00284 (0.003)	-0.00443* (0.002)	-0.00416* (0.003)	-0.00787** (0.004)
East	-0.013*** (0.004)	-0.011*** (0.004)		-0.002 (0.005)
Capital-output ratio	0.039*** (0.006)	0.066*** (0.007)	0.076*** (0.009)	-0.030 (0.085)
Intermediate inputs	0.009 (0.006)	0.070*** (0.009)	0.102*** (0.011)	0.265** (0.123)
Trade union density	0.010** (0.005)	0.007** (0.003)	0.007** (0.003)	-0.029 (0.027)
Intermediate inputs squared	-0.012*** (0.002)	-0.008** (0.004)	-0.006 (0.004)	-0.093* (0.048)
Constant	0.346*** (0.016)	0.269*** (0.013)	0.320*** (0.011)	-0.147 (0.109)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
Limited liability controls	✓	✓	✓	✓
R2	.163	.129	.136	
No. of instruments				114
AR1 (p-value)				.000
AR2 (p-value)				.505
Hansen-J (p-value)				.872
Observations.	36,789	36,789	36,789	27,934

Notes: This table shows results from estimating Equation (20). The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) auto-correlation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

Summarizing, our robustness test by application of microeconomic data and different econometric methods supports our earlier results based on macroeconomic data. Our results are well in line with the empirical literature as, for example, Karabarounis & Neiman (2014) found similar estimates for the firm level labor share. Similar results are also obtained when we drop the years 2003 (Hartz I-II) as well as 2004 (Hartz III) from the analysis as shown in Tables A.3 and A.4.

5.2 Effect heterogeneity with respect to firm size

The narrative of the literature on the labor share is that the decline is primarily driven by large so-called ‘superstar’ firms (e.g. [Autor et al. 2020](#); [Kehrig & Vincent 2021](#)). This interpretation is also consistent with the findings by [De Loecker et al. \(2020\)](#) in which large firms tend to have more power due to less competition and rising markups.

Table 10: Results for ‘large firms’ (> 2,000 employees)

Variables	Labor share			
	OLS (1)	Random effects (2)	Fixed effects (3)	System- GMM (4)
Lag labor share				.423* (.230)
Hartz IV	-.025** (.013)	-.023*** (.009)	-.026*** (.009)	-.053** (.024)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
R ²	.173	.148	.111	
No. of instruments				107
AR1 (p-value)				.006
AR2 (p-value)				.304
Hansen-J (p-value)				.154
Observations	11,940	11,940	11,940	5,709

Notes: This table shows results from estimating Equation (20) for ‘large firms’ > 2,000 employees. The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) autocorrelation are provided by [Arellano & Bond \(1991\)](#). Firm-level clustered standard errors in combination with [Windmeijer \(2005\)](#) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

We test a similar hypothesis and investigate whether the impact of the Hartz IV reform also differs between large and small firms. We choose a value of 2,000 employees as cutoff point, which marks the change from one-third to quasi-parity employee representation in corporate boards (e.g. [Jäger et al. 2021](#); [Gorton & Schmid 2004](#)). Thus, we perform a sample split and estimate the specification for large firms (above 2,000 employees) and small firms (below or equal 2,000 employees) according to our definition. Results for large firms are presented in the following Table 10 and for small firms

in Table 11.

Table 11: Results for ‘small firms’ ($\leq 2,000$ employees)

Variables	Labor share			
	OLS (1)	Random effects (2)	Fixed effects (3)	System- GMM (4)
Lag labor share				.573* (.332)
Hartz IV	.004 (.013)	-.002 (.009)	-.005 (.009)	-.023 (.016)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
R^2	.163	.148	.099	
No. of instruments				103
AR1 (p-value)				.019
AR2 (p-value)				.973
Hansen-J (p-value)				.443
Observations	26,868	26,868	26,868	18,564

Notes: This table shows results from estimating Equation (20) for ‘small firms’ $\leq 2,000$ employees. The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) autocorrelation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

As shown in both tables, there is indeed evidence according to the ‘superstar’ hypothesis as outlined in De Loecker et al. (2020); Autor et al. (2020), that the decline in the labor share is mostly driven by large firms. We find more pronounced effects in Table 10 for large firms compared to the point estimates in Table A.2. Point estimates within each table are also consistent among the estimated OLS, fixed effects as well as System GMM specifications. Our contribution to the literature is that also the firm size, as found by previous literature in another context, is also related to the outside option within a bargaining framework.

6 Conclusion

In this paper we investigate the relevance of bargaining institutions for the decline in the German labor share. The Hartz IV reform, enacted in Germany in the year 2005 provides a unique exogenous reform shock which allows us to estimate the impact of a reduction in the outside option in wage bargaining. Since the reform reduces long-term unemployment benefits for all workers in Germany, the threat point in bargaining between unions and employers is reduced and the threat of unemployment is more severe. We first present a simple bargaining model in which we analyze the relevance of the reduction in the outside option for the decline of the labor share. The model implies that rents are generated within a duopoly in which a union is the bargaining partner. Furthermore, the model connects the wage which is bargained for and the alternative wage which is exogenously reduced because of the Hartz IV reform.

The empirical part consists of three parts. We first combine the EU KLEMS, Penn World Tables and OECD STAN databases to identify the Hartz reforms as a significant structural break in the time series of the labor share. We therefore apply a variety of endogenous change point tests for single as well as multiple breakpoints, in which the tests reveal the Hartz reforms as a significant structural break in the labor share. They additionally point to the fact that, besides the Hartz IV reform, also the reunification of eastern and western Germany is an interesting factor worth considering with respect to the labor share. Second, estimates on the aggregated labor share imply that the Hartz IV reform shock reduces the labor share by around two percentage points, in particular after the first five years after the reform was implemented. Using a synthetic control approach to construct a counterfactual Germany doppelganger, we provide evidence that the effect is rather persistent. In a final robustness section, we additionally use rich firm-level panel data compiled by Bureau van Dijk in combination with fixed effects and System GMM estimation techniques to support the previous findings on the aggregated labor share. We therefore contribute to the burgeoning literature regarding the labor share in the context of technological progress, globalization and mark-ups

stemming from increasing market concentration. In this context, using a unique exogenous reform shock we provide novel evidence of the effects of a reduction in the outside option within a bargaining model and the share of labor.

Besides the endogenously identified Hartz reforms in the labor share series and the application of recent econometric methods in the area of program evaluation, our study is not without limitations. First, there is a debate regarding the measurement of the labor share in which the income of self-employed and real estate income is often neglected (e.g. [Cette et al. 2020](#)). With respect to the database, our data includes income of the self-employed but does not take revenues from real estate into account. Second, because of data limitations, our time series for the construction of the synthetic Germany is slightly shorter compared to the time series which only considers Germany.

With respect to policy implications, we provide a missing link for the effects on the labor share. Whereas studies as, for example, [De Loecker et al. \(2020\)](#) investigate the relevance of decreasing competition and thus an increase in power and mark-ups for the decrease in labor share, we provide evidence from a different direction. Our results show that also the decrease in unionization and thus bargaining power to increase the wage over its marginal product is reduced. Further studies in this context (e.g. [Stansbury & Summers 2020](#)) even argue that the decline in worker power is the major aspect of structural changes in the labor share. For Germany, the decline in unionization and representation of workers by co-determination rights on the establishment level, as mentioned, for example, by [Addison et al. \(2017\)](#), should therefore be in the focus of research on developments of the labor share. In summary, our study adds knowledge to the burgeoning literature on determinants of the labor share. Besides technological progress, globalization and mark-ups, we show within a bargaining framework the relevance of changes in the outside option, which has been neglected so far.

To this end, additional research might be helpful to investigate the reform effect on different types of employment such as temporary or marginal employment. Furthermore, future work should address the question of how to estimate the labor share more

precisely and consistently within and between countries. Since the relevance and interest in technological progress and union power seems likely to increase in the future, more studies are to be expected on this topic.

References

- Abadie, A. (2021). Using Synthetic Controls: Feasibility, Data Requirements, and Methodological Aspects. *Journal of Economic Literature*, 59(2), 391–425.
- Abadie, A., Diamond, A., & Hainmueller, J. (2010). Synthetic control methods for comparative case studies: Estimating the effect of California’s Tobacco control program. *Journal of the American Statistical Association*, 105(490), 493–505.
- Abadie, A., Diamond, A., & Hainmueller, J. (2015). Comparative Politics and the Synthetic Control Method. *American Journal of Political Science*, 59(2), 495–510.
- Acemoglu, D. (2003). Labor- and Capital-Augmenting Technical Change. *Journal of the European Economic Association*, 1(1), 1–37.
- Acemoglu, D., Lelarge, C., & Restrepo, P. (2020). Competing with Robots: Firm-Level Evidence from France. *AEA Papers and Proceedings*, 110, 383–388.
- Acemoglu, D. & Restrepo, P. (2018). The Race between Man and Machine: Implications of Technology for Growth, Factor Shares, and Employment. *American Economic Review*, 108(6), 1488–1542.
- Addison, J. T., Teixeira, P., Pahnke, A., & Bellmann, L. (2017). The demise of a model? The state of collective bargaining and worker representation in Germany. *Economic and Industrial Democracy*, 38(2), 193–234.
- Akyol, M., Neugart, M., & Pichler, S. (2013). Were the Hartz Reforms Responsible for the Improved Performance of the German Labour Market? *Economic Affairs*, 33(1), 34–47.
- Andrews, D. W. (2003). Tests for parameter instability and structural change with unknown change point: A corrigendum. *Econometrica*, 71(1), 395–397.
- Andrews, D. W. K. (1991). Heteroskedasticity and Autocorrelation Consistent Covariance Matrix Estimation. *Econometrica*, 59(3), 817–858.
- Andrews, D. W. K. (1993). Tests for Parameter Instability and Structural Change With Unknown Change Point. *Econometrica*, 61(4), 821–856.
- Andrews, D. W. K. & Ploberger, W. (1994). Optimal Tests when a Nuisance Parameter is Present Only Under the Alternative. *Econometrica*, 62(6), 1383–1414.
- Antoch, J., Hanousek, J., Horváth, L., Hušková, M., & Wang, S. (2019). Structural breaks in panel data: Large number of panels and short length time series. *Econometric Reviews*, 38(7), 828–855.
- Arellano, M. & Bond, S. (1991). Some Tests of Specification for Panel Data: Monte Carlo Evidence and an Application to Employment Equations. *The Review of Economic Studies*, 58(2), 277–297.
- Arif, I. (2021). Productive knowledge, economic sophistication, and labor share. *World Development*, 139, 105303.
- Athey, S. & Imbens, G. W. (2017). The State of Applied Econometrics: Causality and Policy Evaluation. *Journal of Economic Perspectives*, 31(2), 3–32.

- Autor, D., Dorn, D., Katz, L. F., Patterson, C., & Van Reenen, J. (2020). The Fall of the Labor Share and the Rise of Superstar Firms. *The Quarterly Journal of Economics*, 135(2), 645–709.
- Bai, J. & Perron, P. (1998). Estimating and Testing Linear Models with Multiple Structural Changes. *Econometrica*, 66(1), 47–78.
- Bai, J. & Perron, P. (2003). Computation and analysis of multiple structural change models. *Journal of Applied Econometrics*, 18(1), 1–22.
- Barkai, S. (2020). Declining Labor and Capital Shares. *Journal of Finance*, 75(5), 2421–2463.
- Barro, R. J. & Lee, J. W. (2013). A new data set of educational attainment in the world, 1950–2010. *Journal of Development Economics*, 104, 184–198.
- Bazillier, R. & Najman, B. (2017). Labour and Financial Crises: Is Labour Paying the Price of the Crisis? *Comparative Economic Studies*, 59(1), 55–76.
- Benati, L. (2007). Drift and breaks in labor productivity. *Journal of Economic Dynamics and Control*, 31(8), 2847–2877.
- Bental, B. & Demougin, D. (2010). Declining labor shares and bargaining power: An institutional explanation. *Journal of Macroeconomics*, 32(1), 443–456.
- Bentolila, S. & Saint-Paul, G. (2003). Explaining movements in the labor share. *Contributions to Macroeconomics*, 3(1).
- Berthold, N., Fehn, R., & Thode, E. (2002). Falling labor share and rising unemployment: Long-run consequences of institutional shocks? *German Economic Review*, 3(4), 431–459.
- Blanchard, O. & Giavazzi, F. (2003). Macroeconomic Effects of Regulation and Deregulation in Goods and Labor Markets. *The Quarterly Journal of Economics*, 118(3), 879–907.
- Blanchflower, D. G., Oswald, A. J., & Sanfey, P. (1996). Wages, Profits, and Rent-Sharing. *The Quarterly Journal of Economics*, 111(1), 227–251.
- Blundell, R. & Bond, S. (1998). Initial conditions and moment restrictions in dynamic panel data models. *Journal of Econometrics*, 87(1), 115–143.
- Böckerman, P. & Maliranta, M. (2012). Globalization, creative destruction, and labour share change: evidence on the determinants and mechanisms from longitudinal plant-level data. *Oxford Economic Papers*, 64(2), 259–280.
- Botosaru, I. & Ferman, B. (2019). On the role of covariates in the synthetic control method. *The Econometrics Journal*, 22(2), 117–130.
- Bradley, J. & Kügler, A. (2019). Labor market reforms: An evaluation of the Hartz policies in Germany. *European Economic Review*, 113, 108–135.
- Brock, E. & Dobbelaere, S. (2006). Has international trade affected workers' bargaining power? *Review of World Economics*, 142(2), 233–266.
- Caliendo, M. & Kopeinig, S. (2008). Some Practical Guidance for the implementation of propensity score matching. *Journal of Economic Surveys*, 22(1), 31–72.

- Cantore, C., Ferroni, F., & León-Ledesma, M. (2020). The Missing Link: Monetary Policy and The Labor Share. *Journal of the European Economic Association*.
- Card, D. (1992). Using Regional Variation in Wages to Measure the Effects of the Federal Minimum Wage. *ILR Review*, 46(1), 22–37.
- Card, D., Lemieux, T., & Craig Riddell, W. (2020). Unions and wage inequality: The roles of gender, skill and public sector employment. *Canadian Journal of Economics*, 53(1), 140–173.
- Casini, A. & Perron, P. (2019). Structural Breaks in Time Series. In *Oxford Research Encyclopedia of Economics and Finance*. Oxford University Press.
- Cette, G., Koehl, L., & Philippon, T. (2020). Labor share. *Economics Letters*, 188, 108979.
- Chen, Y.-T. (2020). A distributional synthetic control method for policy evaluation. *Journal of Applied Econometrics*, 35(5), 505–525.
- Chow, G. C. (1960). Tests of Equality Between Sets of Coefficients in Two Linear Regressions. *Econometrica*, 28(3), 591–605.
- Ciminelli, G., Duval, R., & Furceri, D. (2020). Employment Protection Deregulation and Labor Shares in Advanced Economies. *The Review of Economics and Statistics*, 1–44.
- De Loecker, J., Eeckhout, J., & Unger, G. (2020). The Rise of Market Power and the Macroeconomic Implications. *The Quarterly Journal of Economics*, 135(2), 561–644.
- Dobbelaere, S. (2004). Estimation of price-cost margins and union bargaining power for Belgian manufacturing. *International Journal of Industrial Organization*, 22(10), 1381–1398.
- Dustmann, C., Fitzenberger, B., Schönberg, U., & Spitz-Oener, A. (2014). From Sick Man of Europe to Economic Superstar: Germany's Resurgent Economy. *Journal of Economic Perspectives*, 28(1), 167–188.
- Eden, M. & Gaggl, P. (2018). On the welfare implications of automation. *Review of Economic Dynamics*, 29, 15–43.
- Elsby, M., Hobijn, B., & Sahin, A. (2013). The Decline of the U.S. Labor Share. *Brookings Papers on Economic Activity*, 2013(2), 1–63.
- Feenstra, R. C., Inklaar, R., & Timmer, M. P. (2015). The Next Generation of the Penn World Table. *American Economic Review*, 105(10), 3150–3182.
- Fichtenbaum, R. (2011). Do Unions Affect Labor's Share of Income: Evidence Using Panel Data. *American Journal of Economics and Sociology*, 70(3), 784–810.
- Fitzenberger, B., Kohn, K., & Lembcke, A. C. (2013). Union Density and Varieties of Coverage: The Anatomy of Union Wage Effects in Germany. *ILR Review*, 66(1), 169–197.
- Giebel, M. & Kraft, K. (2019). External Financing Constraints and Firm Innovation. *The Journal of Industrial Economics*, 67(1), 91–126.
- Gollin, D. (2002). Getting income shares right. *Journal of Political Economy*, 110(2), 458–474.

- Gorton, G. & Schmid, F. A. (2004). Capital, Labor, and the Firm: A Study of German Codetermination. *Journal of the European Economic Association*, 2(5), 863–905.
- Grüner, H. P. (2019). Die politische Zukunft von Hartz IV (The political future of Hartz IV). *ifo Schnelldienst*, 72, 18–21.
- Han, A. K. & Park, D. (1989). Testing for Structural Change in Panel Data: Application to a Study of U.S. Foreign Trade in Manufacturing Goods. *The Review of Economics and Statistics*, 71(1), 135–142.
- Hansen, B. E. (1997). Approximate Asymptotic P Values for Structural-Change Tests. *Journal of Business & Economic Statistics*, 15(1), 60–67.
- Hansen, B. E. (2001). The new econometrics of structural change: Dating breaks in U.S. labor productivity. *Journal of Economic Perspectives*, 15(4), 117–128.
- Hansen, L. P. (1982). Large Sample Properties of Generalized Method of Moments Estimators. *Econometrica*, 50(4), 1029–1054.
- Harju, J., Jäger, S., & Schoefer, B. (2021). Voice at Work. Working Paper 28522, National Bureau of Economic Research.
- Hartung, B., Jung, P., & Kuhn, M. (2018). What Hides behind the German Labor Market Miracle? Unemployment Insurance Reforms and Labor Market Dynamics. *CEPR Discussion Paper No. DP13328*.
- Hochmuth, B., Kohlbrecher, B., Merkl, C., & Gartner, H. (2021). Hartz IV and the decline of German unemployment: A macroeconomic evaluation. *Journal of Economic Dynamics and Control*, 127, 104114.
- Imbens, G. W. & Wooldridge, J. M. (2009). Recent Developments in the Econometrics of Program Evaluation. *Journal of Economic Literature*, 47(1), 5–86.
- Iñaki, E. (2020). The labor share and income inequality: some empirical evidence for the period 1990–2015. *Applied Economic Analysis*, 28(84), 173–195.
- Jäger, S., Schoefer, B., & Heining, J. (2021). Labor in the Boardroom. *The Quarterly Journal of Economics*, 136(2), 669–725.
- Jayachandran, S., Lleras-Muney, A., & Smith, K. V. (2010). Modern medicine and the twentieth century decline in mortality: Evidence on the impact of sulfa drugs. *American Economic Journal: Applied Economics*, 2(2), 118–146.
- Jayadev, A. (2007). Capital account openness and the labour share of income. *Cambridge Journal of Economics*, 31(3), 423–443.
- Kaldor, N. (1957). A Model of Economic Growth. *The Economic Journal*, 67(268), 591–624.
- Karabarbounis, L. & Neiman, B. (2014). The Global Decline of the Labor Share. *The Quarterly Journal of Economics*, 129(1), 61–103.
- Kehrig, M. & Vincent, N. (2021). The Micro-Level Anatomy of the Labor Share Decline. *The Quarterly Journal of Economics*, 136(2), 1031–1087.
- Klinger, S. & Weber, E. (2016). Decomposing Beveridge Curve Dynamics By Correlated Unobserved Components. *Oxford Bulletin of Economics and Statistics*, 78(6), 877–894.

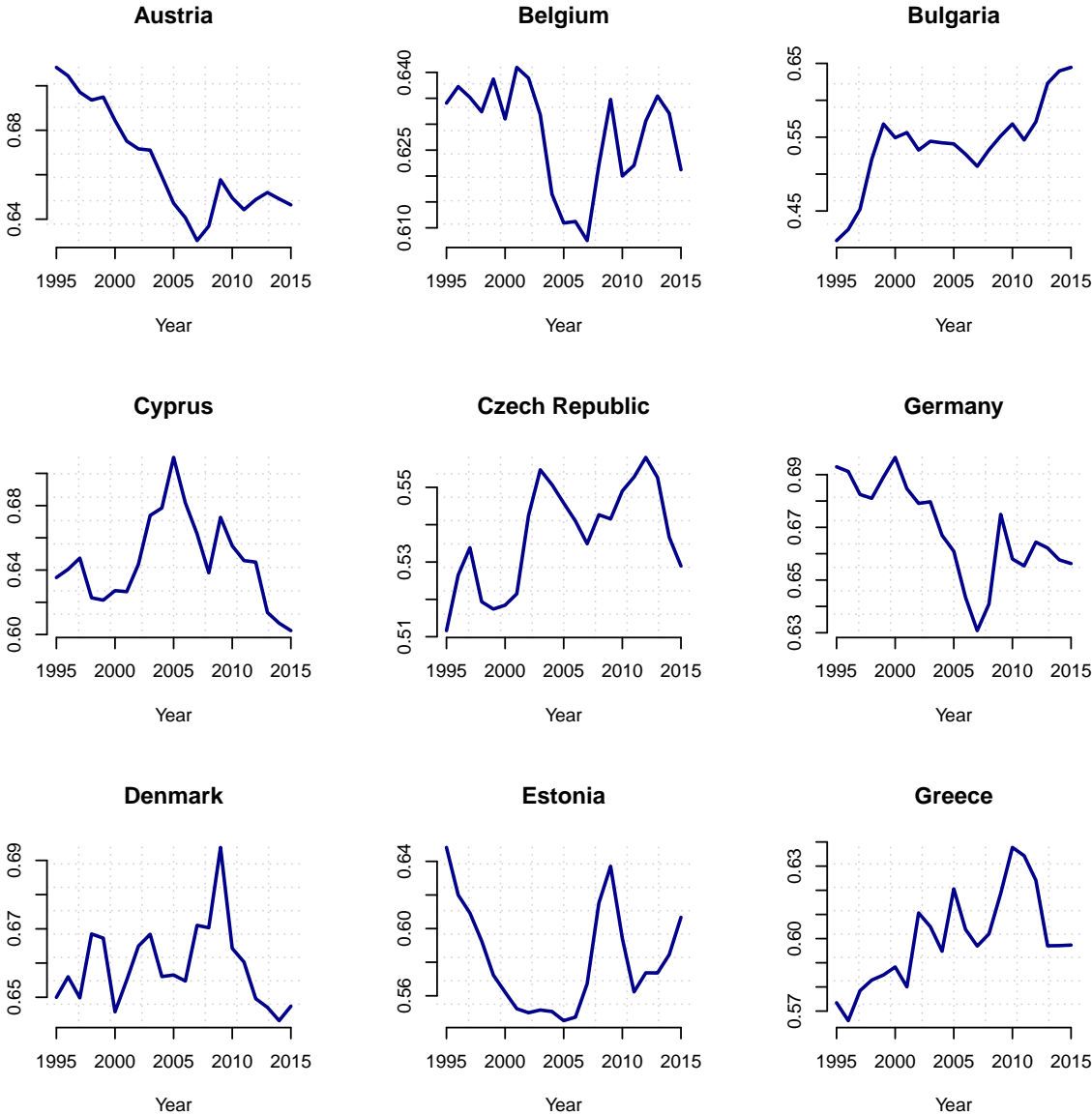
- Knepper, M. (2020). From the Fringe to the Fore: Labor Unions and Employee Compensation. *The Review of Economics and Statistics*, 102(1), 98–112.
- Kraft, K. (1998). The codetermined firm in oligopoly. *Economics Letters*, 61(2), 195–201.
- Kristal, T. & Cohen, Y. (2017). The causes of rising wage inequality: The race between institutions and technology. *Socio-Economic Review*, 15(1), 187–212.
- Launov, A. & Wälde, K. (2016). The employment effect of reforming a public employment agency. *European Economic Review*, 84, 140–164.
- Lee, D. S. & Mas, A. (2012). Long-run impacts of unions on firms: New evidence from financial markets, 1961-1999. *Quarterly Journal of Economics*, 127(1), 333–378.
- Levinsohn, J. & Petrin, A. (2003). Estimating Production Functions Using Inputs to Control for Unobservables. *The Review of Economic Studies*, 70(2), 317–341.
- Link, A. N. & van Hasselt, M. (2019). On the transfer of technology from universities: The impact of the Bayh–Dole Act of 1980 on the institutionalization of university research. *European Economic Review*, 119, 472–481.
- Lunsford, K. G. (2020). Policy Language and Information Effects in the Early Days of Federal Reserve Forward Guidance. *The American Economic Review*, 110(9), 2899–2934.
- Oberfichtner, M. & Schnabel, C. (2019). The German Model of Industrial Relations: (Where) Does It Still Exist? *Jahrbucher fur Nationalokonomie und Statistik*, 239(1), 5–37.
- Onaran, O. (2009). Wage share, globalization and crisis: The case of the manufacturing industry in Korea, Mexico and Turkey. *International Review of Applied Economics*, 23(2), 113–134.
- Peri, G. & Yasenov, V. (2019). The labor market effects of a refugee wave synthetic control method meets the mariel boatlift. *Journal of Human Resources*, 54(2), 267–309.
- Piehl, A. M., Cooper, S. J., Braga, A. A., & Kennedy, D. M. (2003). Testing for structural breaks in the evaluation of programs. *Review of Economics and Statistics*, 85(3), 550–558.
- Piketty, T. & Zucman, G. (2014). Capital is Back: Wealth-Income Ratios in Rich Countries 1700–2010. *The Quarterly Journal of Economics*, 129(3), 1255–1310.
- Psacharopoulos, G. (1994). Returns to investment in education: A global update. *World Development*, 22(9), 1325–1343.
- Quandt, R. E. (1960). Tests of the Hypothesis that a Linear Regression System Obeys Two Separate Regimes. *Journal of the American Statistical Association*, 55(290), 324–330.
- Rodriguez, F. & Jayadev, A. (2013). The Declining Labor Share of Income. *Journal of Globalization and Development*, 3(2), 1–18.
- Rubin, D. B. (1974). Estimating causal effects of treatments in randomized and nonrandomized studies. *Journal of Educational Psychology*, 66(5), 688–701.
- Siegenthaler, M. & Stucki, T. (2015). Dividing the Pie. *ILR Review*, 68(5), 1157–1194.

- Stansbury, B. A. & Summers, L. H. (2020). Declining Worker Power and American Economic Performance. *Brookings Papers on Economic Activity*.
- Stehrer, R., Bykova, A., Jäger, K., Reiter, O., & Schwarzhappel, M. (2019). Industry Level Growth and Productivity Data with Special Focus on Intangible Assets. *Report on methodologies and data construction for the EU KLEMS Release 2019*.
- Stockhammer, E. (2017). Determinants of the Wage Share: A Panel Analysis of Advanced and Developing Economies. *British Journal of Industrial Relations*, 55(1), 3–33.
- Teague, P. & Roche, W. K. (2014). Recessionary bundles: HR practices in the Irish economic crisis. *Human Resource Management Journal*, 24(2), 176–192.
- Wiese, R. (2014). What triggers reforms in OECD countries? Improved reform measurement and evidence from the healthcare sector. *European Journal of Political Economy*, 34, 332–352.
- Windmeijer, F. (2005). A finite sample correction for the variance of linear efficient two-step GMM estimators. *Journal of Econometrics*, 126(1), 25–51.
- Yang, C.-H. & Tsou, M.-W. (2021). Globalization and the labor share in China: Firm-level evidence. *The Manchester School*, 89(1), 1–23.
- Young, A. T. & Zuleta, H. (2018). Do unions increase labor shares? Evidence from US industry-level data. *Eastern Economic Journal*, 44(4), 558–575.

A Appendix

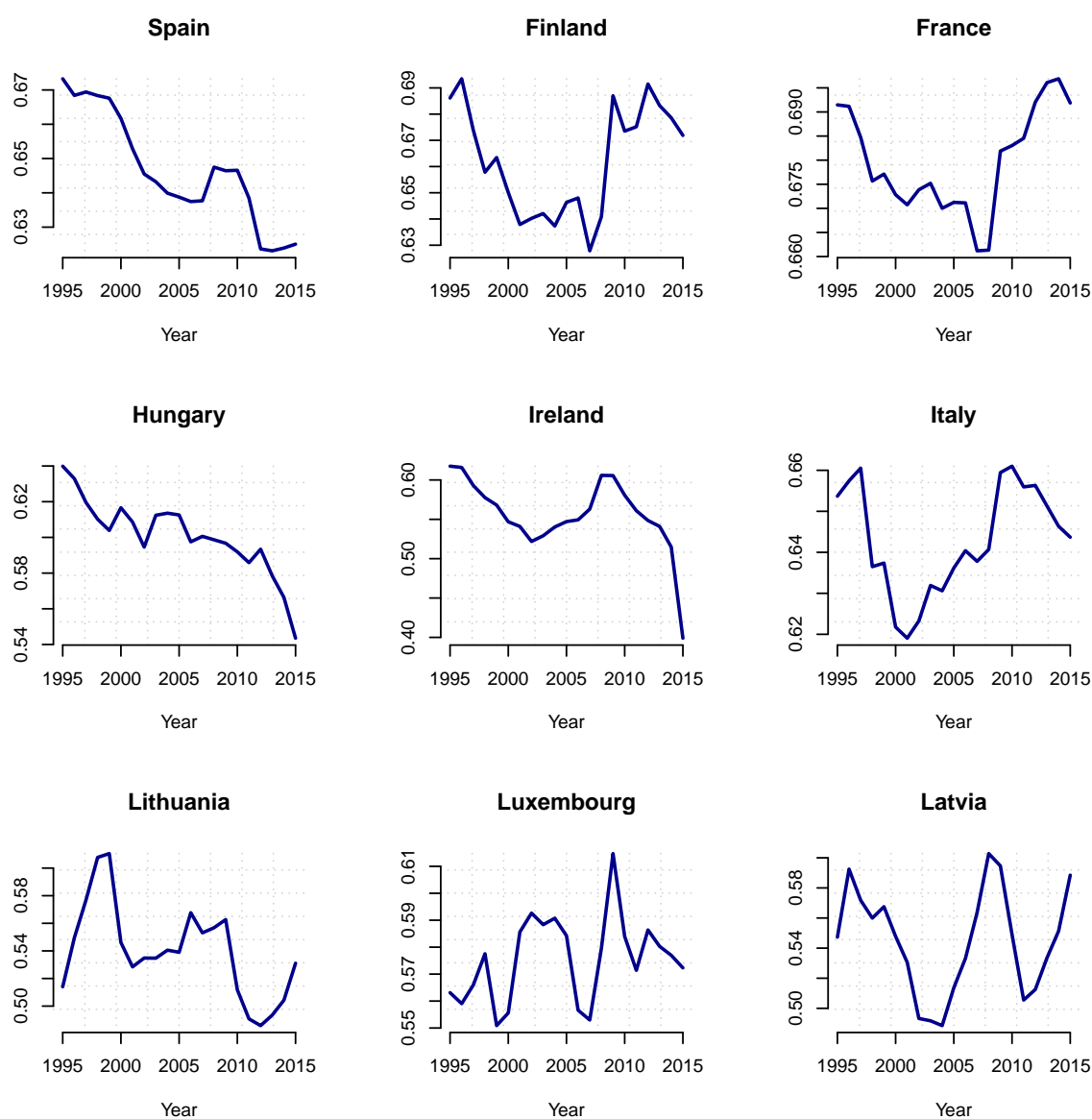
A.1 Figures

Figure A.1: Labor share for different countries



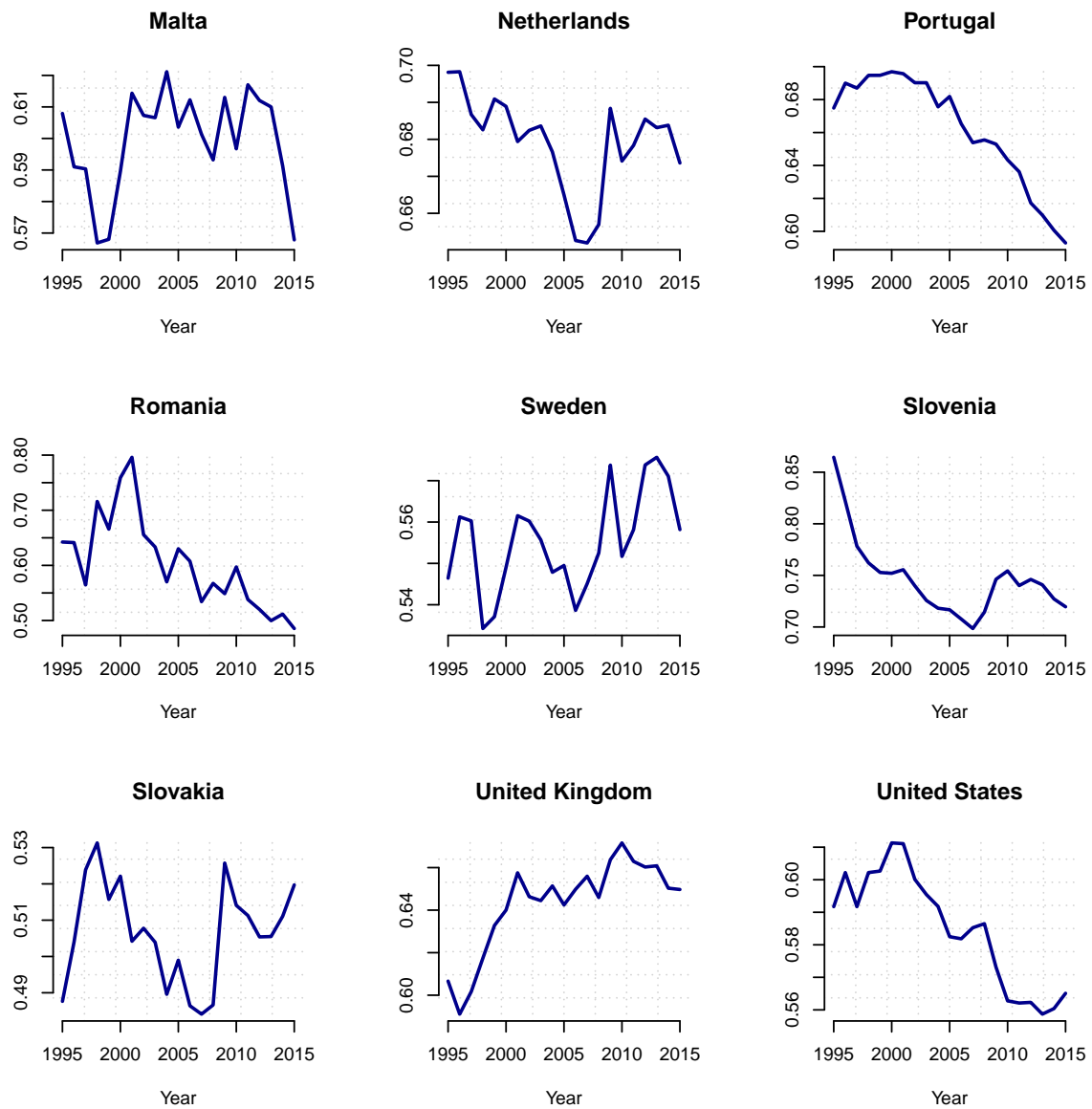
Notes: The figure plots the labor share for different countries between 1995 and 2015. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ which is the total compensation of employees in the economy in relation to gross valued added. Data from the EU KLEMS release 2019. For an overview of variable construction and methodology see [Stehrer et al. \(2019\)](#).

Figure A.2: Labor share for different countries: *continued*



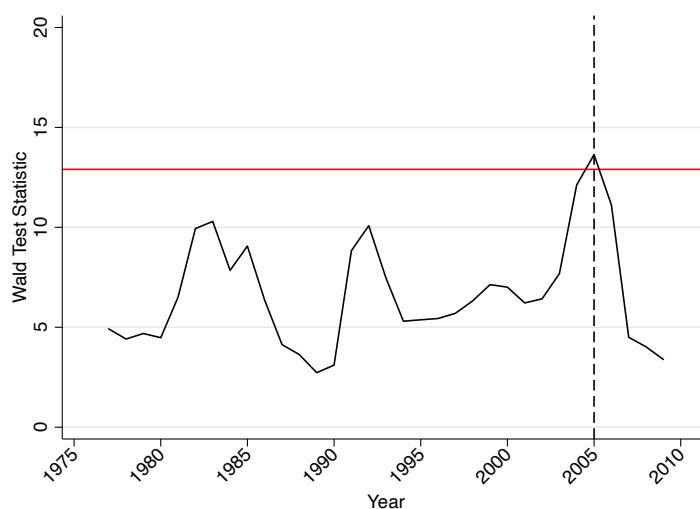
Notes: The figure plots the labor share for different countries between 1995 and 2015. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ which is the total compensation of employees in the economy in relation to gross valued added. Data from the EU KLEMS release 2019. For an overview of variable construction and methodology see [Stehrer et al. \(2019\)](#).

Figure A.3: Labor share for different countries: *continued*



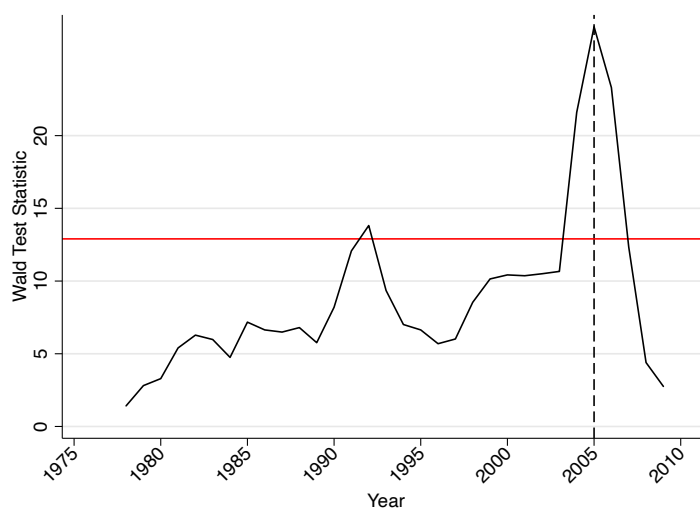
Notes: The figure plots the labor share for different countries between 1995 and 2015. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ which is the total compensation of employees in the economy in relation to gross valued added. Data from the EU KLEMS release 2019. For an overview of variable construction and methodology see [Stehrer et al. \(2019\)](#).

Figure A.4: Wald test robustness I



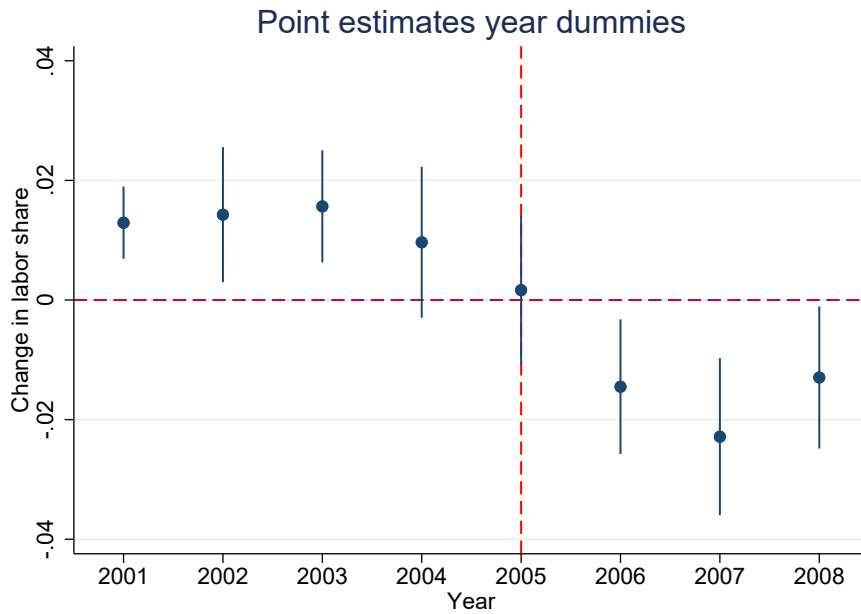
Notes: The figure shows values of the Wald test statistics as outlined in Equations (15) and (16). Regression includes the number of strike days as well as the unemployment rate as additional control variables. Test for a change point in the mean of the labor share (LS_t) in year τ between $t = 1, \dots, T$ where $t = 1977$ and $T = 2009$. The labor share is calculated as outlined in Equation (14). The blue line shows the values of the test statistics. The maximum value (sup Wald) of the Wald test statistic is 13.63 which occurred in 2005, which is exactly the mid-point of the Hartz legislation. The red line indicates the critical value of the test statistics as provided by Andrews (1993, 2003). Values of the test statistics are provided in Table 1. We trim 15 percent from both ends of the sample to ensure that the test has sufficient power.

Figure A.5: Wald test robustness II



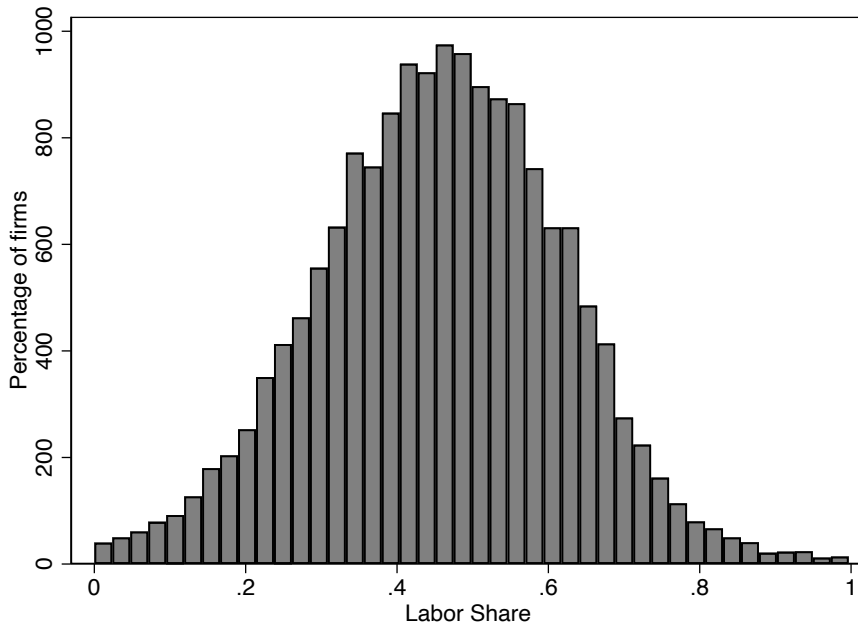
Notes: The figure shows values of the Wald test statistics as outlined in Equations (15) and (16). Regression includes the number of strike days, unemployment rate as well as GDP growth as additional control variables. Test for a change point in the mean of the labor share (LS_t) in year τ between $t = 1, \dots, T$ where $t = 1977$ and $T = 2009$. The labor share is calculated as outlined in Equation (14). The blue line shows the values of the test statistics. The maximum value (sup Wald) of the Wald test statistic is 27.48 which occurred in 2004, which is exactly the mid-point of the Hartz legislation. The red line indicates the critical value of the test statistics as provided by Andrews (1993, 2003). Values of the test statistics are provided in Table 1. We trim 15 percent from both ends of the sample to ensure that the test has sufficient power.

Figure A.6: Hartz IV Dummy variable point estimates



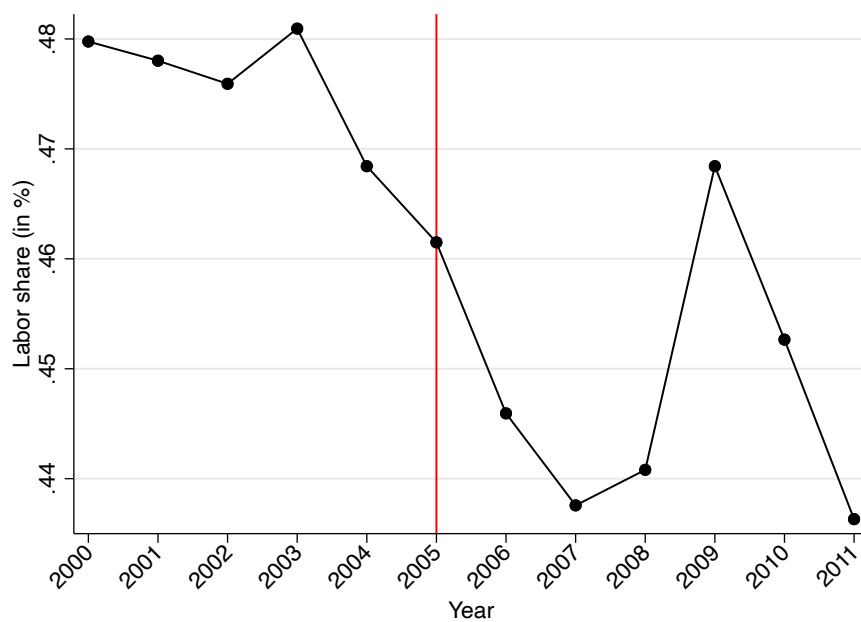
Notes: The figure plots the point estimates from regressing the labor share on all control variables and year dummies as outlined in Equation (18).

Figure A.7: Distribution of the firm-level labor share



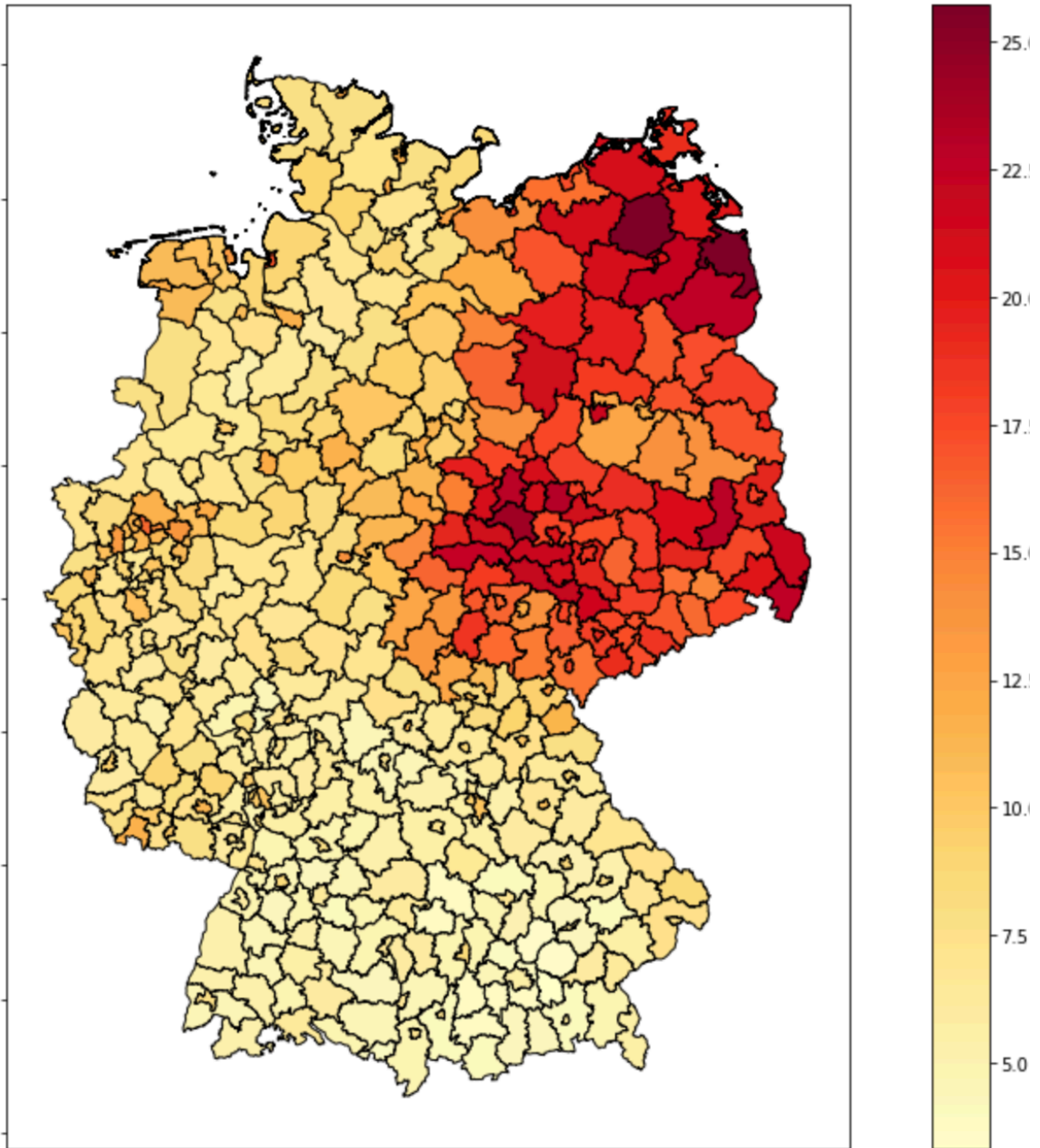
Notes: The figure plots the firm-level labor share. The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011.

Figure A.8: Development of the firm-level labor share



Notes: The figure plots the development of the firm level labor share. The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011.

Figure A.9: German unemployment rates by county in the year 2002



Notes: This figure shows the German county-level unemployment rate in the year 2002. Data from the Federal Employment Agency. The data is matched with firm-level data using the postal code as an identifier for the location of the firm for each county. The unemployment rate is measured in percentage.

A.2 Tables

Table A.1: Description, Explanation and Source of Variables

Variable	Description and Explanation
Dependent Variable	
Labor share	<p><i>Source: EU KLEMS; Bureau van Dijk</i></p> $LS_t = \frac{W_t L_t}{Y_t}$ <p>Which is labor compensation $W_t L_t$ in relation to gross value added Y_t at year t in the corresponding country.</p>
Control Variables	
Share of gross capital formation	<i>Source: Penn World Tables 9.0.</i> The share in relation to GDP measures the stock in real capital within a country.
Share of government consumption	<i>Source: Penn World Tables 9.0.</i> The share in relation to GDP measures the government consumption within a country.
Share of exports	<i>Source: Penn World Tables 9.0.</i> The share in relation to GDP measures the exports within a country.
Share of import	<i>Source: Penn World Tables 9.0.</i> The share in relation to GDP measures the imports within a country.
Human capital index	<i>Source: Penn World Tables 9.0.</i> The human capital index is based on average years of schooling as calculated in Barro & Lee (2013) in combination with an assumed rate of return to education stemming from a Mincer equation as defined in Psacharopoulos (1994) . See also the more comprehensive and detailed definition In the PWT Definition File .
Unemployment rate	<i>Source: The World Bank.</i> Unemployment is the share of the labor force that is without work but available for work and seeking employment.
Trade union density	<i>Source: OECD STAN.</i> Members of trade unions compared to total employment within in defined industry.

Other variables are drawn from the ‘dafne’ dataset compiled by Bureau van Dijk. Description as outlined in Section 5. See Section A.3 for the construction of the union density.

Table A.2: Results for firm-level regressions

Variables	Labor Share			
	OLS (1)	Random effects (2)	Fixed effects (3)	System- GMM (4)
Lag labor share				0.789*** (0.101)
Hartz IV	-0.016* (0.009)	-0.018*** (0.006)	-0.020*** (0.006)	-0.028** (0.012)
Capital-output ratio	0.040*** (0.006)	0.065*** (0.007)	0.075*** (0.009)	0.061 (0.132)
Trade union density	0.010** (0.005)	0.008** (0.003)	0.007** (0.003)	0.039 (0.095)
Intermediate inputs	0.009 (0.006)	0.072*** (0.009)	0.106*** (0.011)	0.526*** (0.155)
Intermediate inputs squared	-0.010*** (0.002)	-0.008** (0.003)	-0.007* (0.004)	-0.192*** (0.056)
Western Germany	0.016*** (0.003)	0.016*** (0.003)		-0.011* (0.006)
Constant	0.352*** (0.016)	0.269*** (0.013)	0.319*** (0.012)	-0.183 (0.267)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
Limited liability controls	✓	✓	✓	✓
R2	.159	.128	.136	
No. of instruments				105
AR1 (p-value)				.000
AR2 (p-value)				.602
Hansen-J (p-value)				.422
Observations.	38,808	38,808	38,808	29,501

Notes: This table shows results from estimating Equation (20). The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) auto-correlation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

Table A.3: Results with unemployment variation, without the years 2003 and 2004

	OLS	RE	FE	Sys-GMM
Lag labor share				0.689*** (0.135)
Unemploy	0.001 (0.001)	0.001 (0.001)		0.008 (0.005)
Hartz IV \times Unemploy	-0.00064** (0.00031)	-0.00080*** (0.00021)	-0.00080*** (0.00021)	-0.00607* (0.00336)
Capital-output ratio	0.040*** (0.006)	0.069*** (0.008)	0.084*** (0.010)	-0.081 (0.160)
Intermediate inputs	0.008 (0.006)	0.069*** (0.009)	0.108*** (0.011)	0.481** (0.188)
Trade union density	0.012** (0.005)	0.010*** (0.003)	0.010*** (0.003)	0.014 (0.041)
Intermediate inputs squared	-0.009*** (0.002)	-0.008** (0.003)	-0.006 (0.004)	-0.169** (0.070)
Western Germany	0.017*** (0.005)	0.019*** (0.006)		0.031 (0.022)
Constant	0.344*** (0.019)	0.263*** (0.016)	0.318*** (0.010)	-0.182 (0.153)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
Limited liability controls	✓	✓	✓	✓
R2	.161	.135	.145	
No. of instruments				110
AR1 (p-value)				.782
AR2 (p-value)				.783
Hansen-J (p-value)				.946
Observations.	31,513	31,513	31,513	24,372

Notes: This table shows results from estimating Equation (20). The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) auto-correlation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

Table A.4: Results with east/west variation, without the years 2003 and 2004

	OLS	RE	FE	Sys-GMM
Lag labor share				0.634*** (0.114)
Hartz IV \times East	-0.004 (0.004)	-0.005* (0.003)	-0.004 (0.003)	-0.023*** (0.009)
East	-0.012*** (0.005)	-0.011*** (0.004)	0.000 (.)	0.009 (0.008)
Capital-output ratio	0.040*** (0.006)	0.070*** (0.008)	0.085*** (0.010)	-0.010 (0.122)
Intermediate inputs	0.008 (0.006)	0.070*** (0.009)	0.108*** (0.012)	0.398** (0.183)
Trade union density	0.013** (0.005)	0.011*** (0.003)	0.010*** (0.004)	-0.008 (0.033)
Intermediate inputs squared	-0.010*** (0.002)	-0.009** (0.004)	-0.007 (0.005)	-0.142** (0.069)
Constant	0.354*** (0.017)	0.279*** (0.014)	0.319*** (0.011)	-0.138 (0.135)
Industry fixed effects	✓	✓	✓	✓
Time fixed effects	✓	✓	✓	✓
Limited liability controls	✓	✓	✓	✓
R2	.160	.133	.144	
No. of instruments				111
AR1 (p-value)				.234
AR2 (p-value)				.223
Hansen-J (p-value)				.941
Observations.	31,513	31,513	31,513	24,372

Notes: This table shows results from estimating Equation (20). The data is based on the “dafne” dataset compiled by Bureau van Dijk; years 2000 – 2011. The labor share is calculated as $LS_t = \frac{W_t L_t}{Y_t}$ as explained in Section 3. The Hansen (1982) test is applied to check for over-identifying restrictions, and tests for first- AR(1) and second-order AR(2) auto-correlation are provided by Arellano & Bond (1991). Firm-level clustered standard errors in combination with Windmeijer (2005) finite sample correction in parentheses. *, **, and *** denote statistical significance at the .1, .05 and .01 level respectively.

A.3 Construction of the industry union density measure

Data. For the construction of the union density measure on the industry level we use the number of trade union members as provided by the [German trade union association](#) as well as the [EU KLEMS](#) data set for information regarding the number of employees in every industry.

Calculation. For the calculation we performed the following steps:

- Trade unions must be assigned to the sectors in order to obtain a specific degree of organization. The constitutions (*Satzungen*) of the DGB trade unions were reviewed and the organizational areas listed were assigned to the respective sectors.
- The degree of organization is recorded as a ratio. This involves dividing the number of union members in a specific union in a year by the number of employees in all industries in a year in which the union is active. A major problem here is that some industries have more than one union. In such cases, the employees in an industry are divided by the number of relevant unions.
- This calculation results in what is known as the gross degree of organization, since union members also include those who have left the labor force.
- The degree of organization calculated for a union as an average across all sectors (relevant for the respective union) is assigned to the sectors in the next step. For example, if the union *IG Metall* is relevant in some sectors for examples 1 and 2, the same degree of organization is found in both sectors. If several unions are represented in these industries, the unweighted average of the n unions is used.
- The gross degree of organization for industry j at time t and union i is then calculated as:

$$Org_{jt}^{gross} = \frac{M_{it}}{\left[\sum_{j=1}^K \left(\frac{1}{n} B_{jt} \right) \right]} \quad (21)$$

where M is the number of union members in a given trade union, B the number of employees in an industry and K the number of industries which are represented by trade union i and finally n is the number of trade unions in a given industry j .