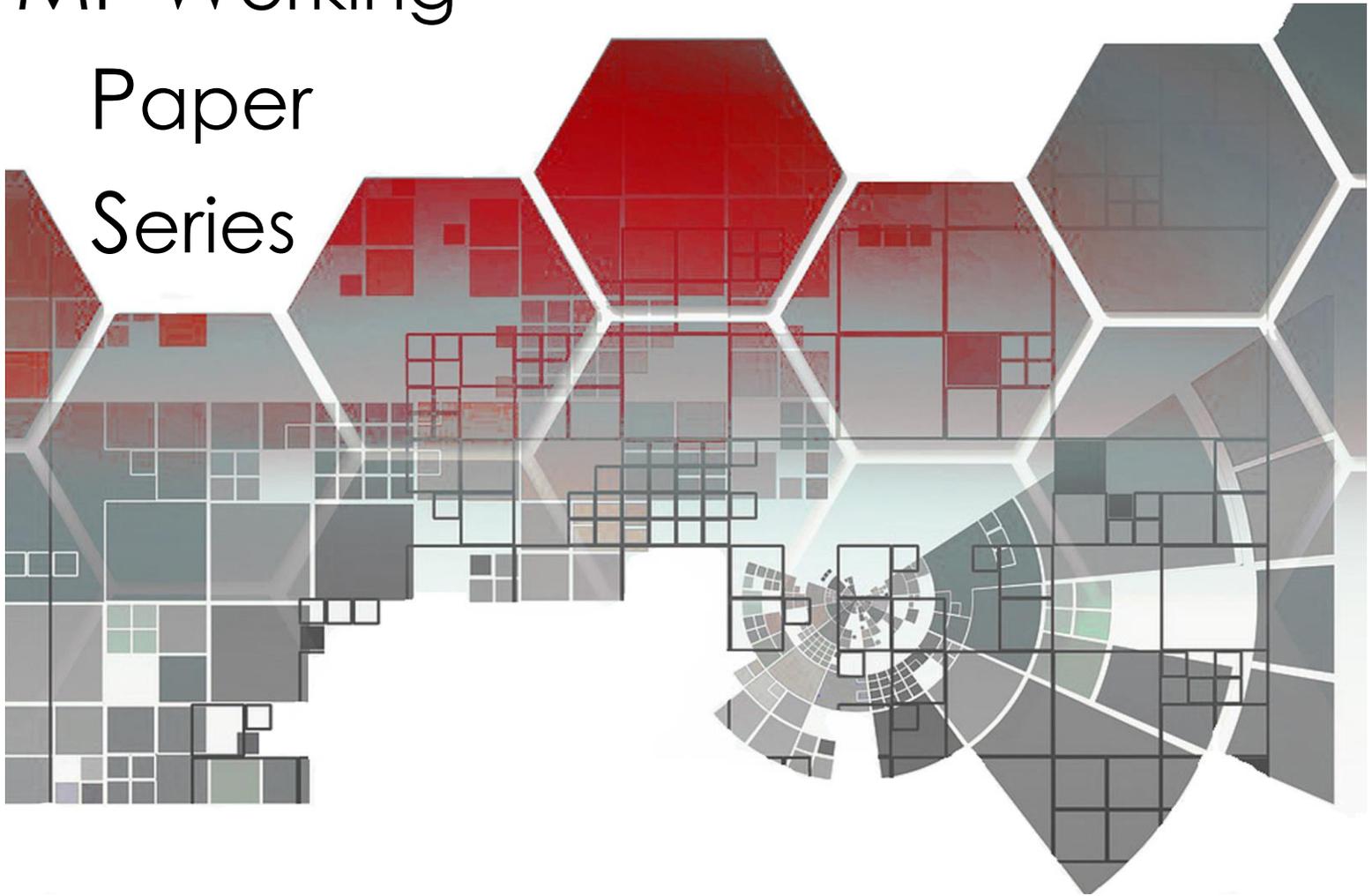


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Labor market externalities of pre-retirement employment
protection

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Using population-level administrative data, we document effects of changes in age-specific employment protection legislation (EPL) on labor market outcomes of those approaching eligibility. Our results show no economically meaningful overall effects of the legislation on employment or earnings of either men or women. Considering separately incumbent workers and non-employees we find small positive and small negative employment effects for the former and the latter groups, respectively. The positive employment effects plausibly reflect additional effort exerted to remain employed up to the coverage threshold.

JEL: J63, J21, J23

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1 Introduction

The last three decades saw sharp declines in fertility and increases in the size of older population in many developed countries. Average total fertility rate in OECD countries declined from 1.98 in 1990 to 1.61 in 2019, while fraction of population aged 65 plus increased from 11.4 to 17.1% over the same time frame. This led to stark increases in old age dependency ratios which are projected to be above 3.6 by the end of the 2020s, meaning that there will be on average 3.6 individuals aged 65 and over per 10 people of working age - at ages 20 to 64 (see panels a and b of Online Appendix Figure A.1).

Given these trends, there is a growing policy interest concerning incentives and regulations aimed at extending working lives and increasing productivity of older workers (Eyster et al. 2008; Deelen and Jongen 2009; OECD 2019; Abraham and Houseman 2020). Although labor force participation of near-retirement men and women has increased in the last three decades, it is still well below the rates of prime age population and has been growing at a sluggish pace since the 1990s (panels c and d in Online Appendix Figure A.1). Job retention rates among these workers tend to be low (Diebold et al. 1997; Hardy et al. 2018) and on average their job search lasts longer before they find a suitable job offer (Faberman and Kudlyak 2019).

On one hand, improved working conditions, changes in the nature of work, and better health status of recently aging cohorts should improve their labor market opportunities. On the other, these factors have not eliminated concerns about the costs of employment of this group of workers from the firm's perspective. The primary risks on the demand side include age-related reductions in productivity, high fixed costs of training, challenges in adjustment to new technologies, and the inability to adjust the wages or hours of such workers downwards in settings with strict employment regulations (see e.g., Lazear (1990), Abowd and Kramarz (2003), Daniel and Heywood (2007), Perek-Białas and Turek (2012), Behaghel et al. (2014)).

To address the risks of insufficient labor demand for older workers, many countries implement policies which make it hard or costly to lay such individuals off (OECD 2019). Yet, while such policies benefit the groups they cover, the natural concerns relate to potential negative labor market externalities on those who remain unprotected. In this paper, we focus on a relatively understudied policy: age-specific employment protection legislation (EPL). Since in our application the regulation in question is one of the most consequential from employer's perspective, we view our analysis as providing an upper bound on the potential effects stemming from more moderate policies protecting older workers.

We ask the following research questions relevant to understanding of the externalities from age-specific employment protection: Are there negative effects of pre-retirement employment

protection on employment and earnings of individuals nearing the eligibility threshold? Do these effects differ for men and women who - in our setting - face different retirement ages and labor force participation rates? And finally, do these effects differ by pre-policy employment status, across the earnings distribution, employer type, and characteristics of the local labor market?

We answer these questions by leveraging administrative data from Poland and a tripple-difference research design generated by a quasi-random change in the EPL eligibility, which was triggered by a retirement age reform. We take advantage of individual-level data from joint social security and tax registers which allow us to track a near-universe of individuals (both employed and not employed) who approach the employment protection eligibility cutoff between January 2015 and June 2018. The exogenous change in the cutoff is generated by a reform passed in November 2016 and implemented in October 2017 which unexpectedly (and unintentionally) granted the EPL to 24 and 27 monthly birth cohorts of men and women, respectively.

We find no economically meaningful effects of the age-specific EPL on employment and earnings of workers nearing the eligibility threshold. Our pooled estimates on a sample of employees and non-employees suggest statistically insignificant, at conventional levels, average employment effects of -0.04 and -0.20 percentage points (pp) for men and women, respectively. Using 95% confidence intervals we can thus rule out negative effect sizes larger than -0.5% and -0.6% for these two groups. Dynamics of earnings, for those employed, are likewise unaffected. Considering incumbent workers and those non-employed prior to the reform separately, we find small positive employment effects for the former and small negative effects for the latter population. Delving further into the heterogeneity analysis, we find that these positive effects are concentrated among workers in the most precarious conditions, those employed in high turnover companies and with lowest earnings. Therefore, it appears that for these workers in particular positive supply side effects dominate any negative labor demand implications. Put differently, these are the workers who could gain the most from exerting extra effort and ensuring that they are not laid off just before reaching the protection threshold.

This paper makes contributions to several strands of the literature. First, we extend the limited set of papers examining labor market consequences of age-specific EPL. [Behaghel et al. \(2008\)](#) show that French 'Delalande tax', which imposed an additional financial cost on employers for terminating contracts of those aged 50+ (in force 1987-2008), led to negative employment effects for older workers. [Saez et al. \(2023\)](#) show that the age 67 removal of EPL in Sweden leads to increased job separations which are particularly pronounced for workers with stronger initial EPL and those plausibly less productive from firm's perspective. Finally, simulated reductions in employment protection for older workers in the Netherlands suggest increases in labor demand for the 55-64 age group ([Deelen and Jongen 2009](#)). Except for these three studies there is an im-

portant knowledge gap recently highlighted in [OECD \(2019\)](#) when it comes to the consequences of age specific EPL. Furthermore, to the best of our knowledge, no prior work examined the potential externalities from EPL coverage for those nearing eligibility - a group of workers which is most likely to be negatively affected.

Second, we add to the broader EPL literature. Cross-country studies show that stringent EPL reduces demand for labor ([Lazear 1990](#); [Kahn 2007](#)), although this association is sensitive to the business-cycle ([Messina and Vallanti 2007](#); [Duval et al. 2020](#)). There is also evidence that restrictive labor legislation lowers productivity growth of industries ([Bassanini et al. 2009](#)) and increases investments in technology ([Griffith and Macartney 2014](#)). On the other hand, [Bassanini and Garnero \(2020\)](#) find no increase in separation rates but rather reduced rates of within-industry transitions while [Kahn \(2010\)](#) finds no employment effects of more lenient EPL. Studies using within-country data and focusing on specific regulations and reforms likewise predominantly find adverse effects of the EPL. [Autor et al. \(2007\)](#), [Kugler and Pica \(2008\)](#) and [Kan and Lin \(2011\)](#) show that stricter EPL reduces employment flows and job turnover. [Hijzen et al. \(2017\)](#) show that Italian firms facing stricter EPL increase their hiring of workers on temporary contracts (uncovered by the EPL). [Boeri and Garibaldi \(2007\)](#), [Sestito and Viviano \(2018\)](#) and [Yoo and Kang \(2012\)](#) document negative relationship between EPL and employment that can, however, be transitory. [Cahuc et al. \(2019\)](#) show that stricter EPL leads to increased separation rates for workers nearing the additional coverage - an example of how EPL could produce negative externalities for uncovered workers. Conversely, [Ichino and Riphahn \(2005\)](#), [Martins \(2009\)](#), [Jacob \(2013\)](#), and [Bjuggren \(2018\)](#) provide evidence that lowering the the level of protection could increase labor productivity and firm performance. When it comes to wages the results are more mixed with [van der Wiel \(2010\)](#) finding increases in wages while [Leonardi and Pica \(2013\)](#) finding decreases as a result of stricter EPL.

Third, we contribute to research on labor demand for workers nearing retirement age. Experimental work documents robust discrimination in hiring of older workers ([Bendick et al. 1997, 1999](#)) with more recent studies highlighting the particular disadvantage of older women ([Lahey 2008](#); [Neumark et al. 2019](#)). Furthermore, [Boockmann et al. \(2012\)](#) and [Huttunen et al. \(2013\)](#) document limited employment effects of wage subsidies and lower payroll taxes for older workers, respectively. On the other hand, [Albanese and Cockx \(2019\)](#) show that such wage subsidies increase retention rates of older workers. Although much of the retirement age policies focus on the supply side (e.g. [Krueger and Pischke \(1992\)](#), [Staubli and Zwimüller \(2013\)](#) or [Laun and Palme \(2023\)](#)), firm's demand for older workers could likewise be affected by such legislation. For example, [Hakola and Uusitalo \(2005\)](#) show that sharing early retirement expenses with employers reduces early labor market exits of older workers in Finland; [Frimmel et al. \(2018\)](#) show that Austrian firms play an active role in the determination of their workers' retirement age ef-

fectively pushing out those that are more costly; [Rabaté \(2019\)](#) reaches similar conclusion using the progressive ban on mandatory retirement in France; and [Morris and Dostie \(2023\)](#) show that banning mandatory retirement in Canada does not lead to changes in demand for older workers. Additionally, [Hairault et al. \(2010\)](#), [Ilmakunnas and Ilmakunnas \(2015\)](#) and [Bertoni and Brunello \(2021\)](#) study how retirement age policies affect workers prior to the retirement eligibility. They all suggest that workers near the eligibility cutoff suffer in terms of their employment prospects - a result prompting potential concerns of negative spillovers of the EPL for those approaching eligibility. Finally, although this perception is not necessarily supported by the data, one of the main concerns when it comes to the demand side is declining health and productivity of older workers. In fact, [Mahlberg et al. \(2013\)](#) document that productivity of Austrian firms is not related to the share of older workers they employ while [Börsch-Supan and Weiss \(2016\)](#) show that individual worker productivity in an assembly plant is stable until at least age 60. One reason for the flat age-productivity gradient could be technological change which may disproportionately benefit older workers, especially in physically demanding jobs ([Gordo and Skirbekk 2013](#)).

Finally, we contribute to research on societal aging and its consequences. The aforementioned increases in old age dependency ratios put strain on the solvency of the social security systems (see [Jimeno et al. \(2008\)](#) for overview of this research) and will likely lead to lower economic growth ([Kotschy and Bloom 2023](#)). In particular, [Kitao \(2014\)](#) suggest that making US Social Security sustainable given the current demographic trends would require drastic changes in either taxation or generosity of the benefits. Furthermore, restricted employment opportunities of older workers could increase inequality and poverty rates among the elderly ([Deaton and Paxson 1998](#)). These factors could be one reason behind increasing voluntary retirement age in the US ([Brown et al. 2022](#)). Finally, recent work suggests that rapid population aging will soon become an important policy consideration beyond the developed world including in Africa ([Duhon et al. 2023](#)).

We view our results as having two key policy implications. First, we show that age-related employment protection does not have major negative externalities with regard to labor demand effects for the soon-to-be-covered individuals. Since the protected workers are guaranteed employment until the retirement age, unless they decide to quit themselves, without the negative externalities the policy actually increases aggregate employment rates of older workers. Thus, our findings question the common policy concern that pre-retirement EPL leads to a trade-off between benefiting those already protected and hurting those employees who are nearing eligibility. Second, our results highlight the importance of considering incumbent and not employed workers separately. We show that the null overall effect stems from slight positive employment effects for the incumbents and slight negative employment effects for those who are not employed. The implications for the latter group, despite relatively small effect sizes of the estimates,

could be of a concern for policy. The positive effects for incumbent workers are further exacerbated among those working in more precarious jobs - those lower paid and employed in firms with high turnover - which is consistent with the fact that EPL could induce increases in effort of some workers who are trying to avoid being laid off in the final months without the coverage.

2 Institutional background

2.1 Pre-retirement employment protection legislation

Pre-retirement employment protection legislation (EPL) in Poland is regulated by the Labor Code, a set of laws which is separate from the Civil Code and focuses solely on labor relations. Article 39 of the Code specifies that an employer cannot terminate a contract if an employee has at most 4 years left until the retirement age and is eligible to retire at that time. This regulation has been present in the Polish Labor Code since its inception in June 1974, though it was extended from two to four years in 2008.¹ It applies to both open-ended and fixed-term labor contracts but excludes those who are eligible for disability pensions. If the company breaks the law and dismisses an employee who is eligible for employment protection it can be sued in the court on the basis of both the Labor Code and the Criminal Code. Importantly, the regulations encompass not only inability to terminate a contract but also lowering of wages, changing of hours of work, or moving to a more burdensome role or position. Exceptions include mass layoffs, dismissal due to disciplinary reasons, or restructuring of the wage scale for the whole company, however, these are not easily gameable by employers and are monitored closely by the authorities. Overall the Polish pre-retirement EPL is inflexible, offers little leeway to employers and thus could be costly from their perspective. The aggregate statistics suggest that the law is broadly obeyed while the courts tend to side with plaintiffs when it comes to its violations.

2.2 Retirement age legislation

Retirement regulations prior to January 1st, 2013 stipulated a retirement age of 65 and 60 years for men and women, respectively. This was changed by the 2012 reform which legislated a gradual increase in the statutory retirement age so that it becomes equalized at age 67 for both genders. Specifically, the retirement age was to grow gradually beginning in January 1st 2013, with increases of one month per each calendar quarter. Thus, men would have reached their

¹The extension of the protection period from 2 to 4 years in 2008 was introduced as part of an agreement between the government and social partners in a package limiting access to early retirement. In 2023 an additional regulation was implemented further strengthening protection by forcing the employers to cover employees' wages during the potential litigation related to unlawful dismissal.

target age of 67 in October 2020 while women in October 2040, due to the initial five year gap between the two groups.

The 2012 reform, which had been unannounced in the electoral campaign of 2011, proved unpopular and the return to the pre-reform retirement age became one of the key pledges of the opposition candidate, Andrzej Duda, in the 2015 presidential race. The promise played an important role in Duda's electoral victory and the theme of returning to a lower retirement age was again featured prominently in the parliamentary elections in October 2015 (see Google search statistics presented in Figure 1). The incumbent coalition parties ended up losing both the presidential and the parliamentary elections and before the end of the year the new president presented the legislation to the parliament, which stipulated reverting to retirement age regulations from before 2013. This was done despite rapid population aging and against the economic and policy analyses provided both by independent entities and by the government itself, who was far from enthusiastic to embrace the change.

Following a freeze on the parliamentary discussion on this issue and a series of alternative options which were considered by the government, in the end the presidential proposal returned to parliament a year later. After a very brief debate in the parliament it was swiftly passed in November 2016 and signed into law on December 19th 2016 (Google search statistics presented in Figure 1 confirm increased interest in the issue of retirement age at the time). The new regulations came into effect on October 1st 2017, after less than a year of a hold-up period. By the time the new law came into force, multiple cohorts saw their retirement age had grown as a result of the gradual increases implemented since 2013. A month earlier, in September 2017, retirement eligibility was granted to men born in July 1951 (i.e. aged 66 and 2 months) and to women born in July 1956 (i.e. aged 61 and 2 months). The reversal meant that all cohorts who were 65 (men) or 60 (women) on October 1st 2017 were granted retirement rights on that day. Unsurprisingly, the reform attracted a lot of attention and thousands of individuals took advantage of the opportunity to retire in accordance with the new regulations. In the last quarter of 2017 and first quarter of 2018, Polish Social Security Institution (ZUS) registered over 357 thousand new pension claims, while in 2015 and 2016 the same two quarters saw just over 100 thousand new claims.

2.3 Using the 2016 reform to estimate the effects of pre-retirement employment protection on workers approaching eligibility

Since the EPL was not defined with respect to a specific age, but rather with respect to retirement age, cohorts whose retirement age grew after the 2012 reform became eligible for the benefit at a later age. Conversely, the policy reversal automatically (and unexpectedly) made multiple younger cohorts eligible for the EPL. Importantly, discussions surrounding the 2016 reform rever-

sal clearly drew attention primarily to the statutory retirement age and not the implied extension of the EPL (see Google search statistics in Figure 1). And yet, the gradual increase in retirement age for subsequent cohorts, meant that there were 24 monthly birth cohorts of men and 27 cohorts of women, who on October 1st, 2017 became covered by the additional age-related employment protection benefits.² Critically from the perspective of our research design, the workers who were about to gain the EPL coverage on October 1st 2017 were nonetheless unprotected during the policy hold up period, i.e. the time from the date it was signed into law (i.e., December 19th 2016) to when it was implemented (i.e., October 1st, 2017). This is the period in which the employers (and employees) could adjust their employment strategies with the new rules in mind.

Details on how the reforms affected specific cohorts are illustrated in Figure 2 which maps the two retirement age regimes in two dimensions: the month of birth and the calendar month (separately for men - Figure 2a, and women - Figure 2b). The Figure shows the dynamics over time (horizontal axis) of the 2012 and 2016 retirement age reforms and the corresponding coverage of EPL eligibility by monthly birth cohorts (vertical axis). The gray area delineates the statutory retirement age and reflects eligibility for retirement pension benefits according to the 2012 regulations and the growing retirement age. Retirement age of 67 would have been reached by October 2020 for men and October 2040 for women (not shown on the Figure). Corresponding to the gray area is the red area which reflects the time when specific monthly birth cohorts are eligible for additional age-related employment protection, starting four years prior to reaching retirement age for a specific cohort. The navy-blue shaded area (which overlaps part of the red area) shows additional months of retirement eligibility gained as a result of the 2016 reduction in retirement age which came into force on October 1st 2017. In October 2017 cohorts of men born between August 1951 and October 1952, as well as women born between August 1956 and October 1957 became eligible to claim retirement pensions, in addition to those born in July 1951/56 who would have been the only cohorts gaining retirement pension eligibility according to previous regulations. As it is clear from Figure 2 the retirement eligibility gains were higher for later cohorts, whose retirement age was higher under the 2012 regulations. The orange shaded areas in turn show the additional months of EPL coverage for the younger cohorts for whom the 2016 reform changed the statutory retirement age.

All cohorts of men born after November 1st 1954 and women born after August 1st 1959 would at some point become eligible for the additional employment protection, which, as a result of reverting to the retirement age of 65 and 60, would now cover them as they reach the age of 61 and 56, respectively. In this paper though, we restrict our attention to the 24 monthly cohorts

²The difference in the number of eligible cohorts of men and women results from the fact that the youngest cohorts of men covered by the additional protection in October 2017 would have retired after October 2020 having reached the statutory retirement age of 67.

of men and 27 monthly cohorts of women who became eligible for additional EPL on October 1st 2017. These include men born between November 1st 1954 and October 31st 1956 as well as women born between August 1st 1959 and October 31st 1961. Among those cohorts we select individuals that gained the largest number of months of additional protection, and - in theory at least - would have been of particular concern for their employers. Thus, in our main analysis we define the treatment group as ten monthly cohorts of men and women, born respectively January-October 1956 and January-October 1961. In Section 6 we show, however, that our results remain broadly unchanged when we alter the number of months considered when defining treatment.

To define the control groups, for reasons outlined below, we use cohorts who are two years younger. Although they will eventually also become eligible for the EPL at an earlier age, we assume that in October 2017 their eligibility is far enough into the future that it would not be considered as a factor in their employers' demand decisions. Specifically, we use January-October 1958 cohorts for men and January-October 1963 cohorts for women as controls. We use the same set of months to account for any season-of-birth effects. It is worth noting though, that to the extent that individuals from the control groups are potentially also negatively affected, our estimates should be treated as a lower bound. Additional samples used for robustness analyses are described in more details in Section 6 while details of the samples are presented in Tables A.1 and A.2 in the Appendix.

The choice of cohorts forms the basis of our quasi-experimental design. The first difference is the before-after 2016 reform while the second difference is the 1956(61) vs. 1958(63) birth cohorts for men (women). This yields a standard difference-in-differences design. A complicating factor in this approach, however, is that it ignores any potential age-related differences in labor market outcomes between our treatment and control groups. In other words, even if parallel trends hold in the pre-treatment period, there may be factors related specifically to age which would affect the treated (older) cohorts in the post-period but that do not affect (to the same extent) the control (younger) groups. The most obvious reason in the context of employment of older people is the correlation between age and health: If age-specific effects exist, they could affect the treated and control groups differentially and could invalidate our parallel trends assumption. Thus, to account for the potential age-specific effects we define a third difference where we use cohorts lagged by one year. In essence therefore, we assume that these age-related confounders would be the same for the labor market dynamics of cohorts 1956-58 (1961-63 for women) and 1955-57 (1960-62). In Section 4 we outline how these considerations translate into estimating equations.

3 Data

We use a dataset of combined individual-level information from several administrative sources matching monthly labor market information from the social and health insurance register (ZUS), annual income tax data, and basic individual characteristics (gender, date of birth and death) from the PESEL registry. The data has been compiled at the Polish Ministry of Finance and the matched information begins from January 1st 2015. Labor market data used for our analysis covers the period until June 31st 2018, but we also use data up to October 2020 to specify our sample selection criteria. We therefore observe labor market outcomes for up to 22 months prior to the reform legislation in November 2016, 11 months of the hold up period and 9 months after it came into force. The ZUS data include monthly information on all income sources for which the social security and health insurance contributions are paid and have a near-universal coverage of the working population.³ The dataset also facilitates a match between employers and employees as well as includes employer industry codes. We can thus derive such firm-level characteristics as total employment, firm-level turnover and age composition of the workforce at the firm level, all measured prior to the policy change, which we take advantage of in the heterogeneity analysis. On the basis of industry codes we further exclude individuals eligible for industry-specific retirement regulations and employees in the public sector.⁴ Additionally, using data on pension claims from the ZUS registry we drop men/women who by the age of 62/57 claimed either early retirement (so-called “bridge pension”) or disability pensions.⁵ The “bridge pensions” should be thought of as exogenous given that the eligibility for this specific pension, which varies at birth cohort-by-age-by-occupation/industry level, was determined long prior to the reform and applies only to people who for at least 15 years of their career (including time before 1999) worked in “special conditions”.⁶ We drop these individuals since they could confound the reform variation which is partially determined at cohort-by-age level. In robustness analysis we adjust these selection criteria to remove only those claiming “bridge pensions” (i.e. including those who end up claiming disability pensions, in the case of which there is a higher risk of potential endogeneity with respect to the EPL eligibility) as well as broadening them to cover claims of any form of pension. We also change the pension qualification cut-off age to 61.5/56.5 - none of these changes

³The data do not include farmers, students’ with only temporary jobs, and individuals with only result-based contracts. Additionally, we exclude uniformed services as well as judges and public prosecutors from the data because these jobs are regulated separately. A worker can have several records each month (e.g., if they change jobs or work at more than one institution) and in such cases we select the observation within a month with the highest earnings.

⁴Specifically, we exclude the following sectors: primary sector e.g., hunting (A), mining (B), water and sewerage management (E), scientific contracting and research (M), public administration and military (O), education and teaching (P), household production for household use purposes (T), and international organizations (U).

⁵For this purpose we use data from the registry all the way up to October 2020, which is the month in which the youngest control cohort (men born in October 1958, women born in October 1963) turn 62/57.

⁶The list of these conditions includes for example jobs performed underground, heavily demanding physical jobs and those requiring special psycho-physical agility.

have any substantial implications for the overall findings.⁷

The dataset includes all individuals who at any point between 2015-2020 have paid any social or health insurance contributions (including contributions paid by the government’s labor office for the registered unemployed) and/or filed any income tax.⁸ Overall, the starting sample for our main analysis among cohorts considered as treated (i.e. men/women born in January-October, 1956/61) corresponds to 155,0787 men and 150,368 women. After applying the pensions and industry/sector selection criteria we end up with 54,397 men and 60,374 women in the treated sample. Corresponding numbers for the control sample as well as for samples used in the triple difference estimation under various sampling conditions are given in Tables A.1 and A.2, respectively for men and women, in the Online Appendix. The triple difference estimation is based on a sample of over 233 thousand men and 246 thousand women. In the case of estimations in which employment is the outcome variable we use balanced panels, with the final number of individual observations corresponding to the product of individuals and observation months. Earnings equations are based on unbalanced panels of observed earnings (see Online Appendix Tables A.1 and A.2) provided that we observe at least one earnings record in the pre- and the post-reform periods.

We consider two outcomes of interest: an indicator for being an employee and labor earnings. The former variable takes value of one for everyone who is employed on labor contract (“umowa o pracę”), i.e. a contract that is subject to coverage of the EPL, and zero for all other individuals. This means that the non-employed group in our sample includes individuals who are: currently not working, employed on civil contracts, or self-employed.⁹ Labor earnings are monthly earnings on labor contracts as recorded in the social insurance database.

4 Empirical Specifications and Identification

Mapping the institutional setting to the data, recall first that we have monthly level information on labor market outcomes between January 2015 and June 2018. We index these observations by t which runs from 1 (January 2015) to 42 (June 2018). We also have monthly birth cohorts c

⁷Because information on pension receipt (including the type of pension) in the matched administrative dataset is available only from 2016, when we impose the pension claim criterion at 61.5/56.5 we have to limit the monthly birth cohorts we examine to July-October (rather than January-October as in the main specification). This is because the July cohort is the first for which we have pensions observation at the point when the oldest cohorts turn 61.5/56.5 (men born in July 1955, women born in July 1960).

⁸Although registering as unemployed with the labor office is voluntary, it is tied to health insurance coverage and thus most of the unemployed register to receive (temporary) unemployment benefits and (permanent) health insurance coverage.

⁹The latter two groups are not eligible for the EPL coverage. We include these individuals in the main specification because laid off individuals could move to these forms of activity and because firms could change the form of employment to avoid the potential costs of employing individuals covered by the EPL.

which will be differentially affected by the reform changing the eligibility for the EPL. We consider men born in 1956 (1961 for women) as a treatment group while those born in 1958 (1963 for women) as a control group. Since the reform was passed in the parliament in November 2016, we treat this month as the first treated time period (month 23 in our time index t) i.e., $\text{Post}_{it} = \mathbb{1}[t \geq 23]$. These two differences constitute our baseline difference-in-differences setting: we compare the 1956 (1961) monthly birth cohorts to the 1958 (1963) monthly birth cohorts, before vs. after November 2016. The event study for estimating these effects is given by:

$$Y_{ict} = \alpha + \sum_{j=-22}^{-2} \beta_j^1 T_{ic} \mathbb{1}[t - \text{Nov2016} = j] + \sum_{j=0}^{19} \beta_j^2 T_{ic} \mathbb{1}[t - \text{Nov2016} = j] + \gamma_t + \delta_c + \varepsilon_{ict} \quad (1)$$

where Y_{ict} is an indicator equal to one if individual i born in monthly cohort group c observed in calendar month-year t is employed on the EPL eligible contract, and zero otherwise. Additionally, for those employed on EPL eligible contracts we use monthly labor earnings as another outcome variable. T_{ic} is an indicator variable taking value one for men born in January to October 1956 and women born in 1961 (versus zero for men born in January to October 1958 and women born in 1963) while $\mathbb{1}[t - \text{Nov2016} = j]$ are event time dummies running from -22 (t month 1 i.e., January 2015 minus t month 23 i.e., November 2016) to 19 (t month 42 i.e., June 2018 minus t month 23 i.e., November 2016) and we omit October 2016 (indexed as $j = -1$ at t month 22) as the reference period. We also include two sets of fixed effects defining the comparisons: γ_t for calendar month and δ_c for monthly birth cohort. Parameters of interest in Equation 1 are β_j , where β_j^1 represent pre-treatment period estimates allowing examination of the parallel trends assumption and β_j^2 represent post-reform dynamic treatment effects.

One issue with the empirical approach outlined above is that it does not account for differential effects of ageing between the treated and control cohorts. In that, our treatment group is always older, and thus, through health and family shocks, could have adverse labor market outcomes irrespective of the stricter EPL (potentially downward biasing our results). Note that this problem is independent of the parallel trends assumption, which needs to hold in the post-reform period absent the treatment, and it still could be that the post-reform effects are exacerbated due to the fact that the treated cohorts are relatively older compared to the control cohorts. To mitigate this, assuming that the aging effects are the same for cohorts born a year earlier, we introduce

a triple difference event study design of the following form:

$$\begin{aligned}
Y_{ict} = & \alpha + \sum_{\substack{j=-10 \\ j \neq -1}}^{19} \zeta_j D_{ic} \mathbb{1}[t - \text{Nov}2016 = j] + \sum_{\substack{j=-10 \\ j \neq -1}}^{19} \phi_j A_{ic} \mathbb{1}[t - \text{Nov}2015 = j] \\
& + \sum_{j=-10}^{-2} \theta_j^1 (D_{ic} \mathbb{1}[t - \text{Nov}2016 = j]) (A_{ic} \mathbb{1}[t - \text{Nov}2015 = j]) \\
& + \sum_{j=0}^{19} \theta_j^2 (D_{ic} \mathbb{1}[t - \text{Nov}2016 = j]) (A_{ic} \mathbb{1}[t - \text{Nov}2015 = j]) \\
& + \gamma_t + \delta_c + \varepsilon_{ict}
\end{aligned} \tag{2}$$

where Y_{ict} , γ_t , and δ_c are defined as in Equation 1 while we shift the treatment and control group by 12 months to generate the third difference proxying for the aging effect. This is reflected by the fact that in our “true reform” sample the first treatment month remains November 2016 while in our “placebo reform” sample the first treatment month is November 2015. Since our data do not extend before January 2015, due to taking the 12-months lag, we need to trim the left hand side of the event study and hence in this specification $-10 \leq j \leq 19$. We also need to modify the definition of treatment and control group (T_{ic} in Equation 1) and hence we rewrite it as D_{ic} which takes value one for individuals born January to October 1956 (T_{ic} in Equation 1) or January to October 1955 (placebo treatment cohorts) for men and in 1961 or 1960 for women. An indicator variable, A_{ic} , then takes value one for men born January to October 1956 or January to October 1958 (1961 or 1963 for women) reflecting the “true reform” sample cohorts. Parameters of interest in Equation 2 are θ_j , where θ_j^1 represent pre-treatment period estimates allowing examination of the parallel trends assumption and θ_j^2 represent post-reform dynamic treatment effects. In Section 5 we present results from Equations 1 and 2 in parallel on the same set of graphs.

Finally, we present aggregate average treatment effects by estimating the following triple difference equation:

$$\begin{aligned}
Y_{ict} = & \alpha + \theta \text{Post}_{it} \times D_{ic} \times A_{ic} + \zeta \text{Post}_{it} \times D_{ic} \\
& + \phi \text{Post}_{it} \times A_{ic} + \gamma_t + \delta_c + \varepsilon_{ict}.
\end{aligned} \tag{3}$$

Recall that the reform was passed by the parliament in November 2016, signed into law in December 2016, but became binding only on October 1st 2017. Thus, employers could react to the reform by adjusting their employment all the way up to September 2017 and we might ex-

pect the spillover effects of the additional protection for our treated cohorts to be most evident close to and beyond the time when the retirement age reform comes into force. Note, however, that as time passes eventually the control cohorts in our design also begin to get closer to their own age-specific protection. In this case, if external effects of approaching protection are homogeneous across cohorts, the difference between our treated and control cohorts could be attenuated. To account for this, when considering pooled regressions, the estimation is based on data for six pre- and six post-treatment months. Pre-treatment months include May-October (2016 for “true reform” and 2015 for “placebo reform”), while post-treatment months, defining $Post_{it}$ in Equation 3, cover September-February (2017-2018 for “true reform” and 2016-2017 for “placebo reform”). Other than that Y_{ict} , D_{ic} , A_{ic} , γ_t , and δ_c in Equation 3 are defined as in Equation 2. The parameter of interest in Equation 3 is θ and it describes the average treatment effect of stricter EPL on labor market outcomes of workers approaching the eligibility threshold compared to workers father away from eligibility, net of the aging effect accounted for by the third difference. Since we have repeated observations on the same individuals, in all estimating equations we cluster the error term, ε_{ict} , at the individual level.

5 Results

Figure 3 presents our main results for men (left column) and women (right column) in the form of event studies as specified in Equations 1 and 2. The dashed vertical line defines October 2016, the last month before the reduced retirement age was legislated in the parliament, while the solid vertical line represents October 2017, when the legislation become binding and workers in the treatment group became covered by additional age-related employment protection. Thus, the period between November 2016 and October 2017 is our hold-up period where we expect most of the adjustment to the forthcoming protection eligibility - both on the employer and the employee side - to happen. The top row (Figures 3a and 3b) shows overall estimates for employment, while the middle row (Figures 3c and 3d) the estimates for earnings conditional on employment. The gray plots represent the difference-in-differences event studies (Equation 1) and the black plots show the triple difference estimates (Equation 2). Considering the former design we see no pre-trends when it comes to employment but a declining trend in the pre-reform period for earnings. The latter reflects our concerns related to the fact that our treatment group is two-years older than our control group, and may therefore experience different labor market outcomes regardless of any labor market legislation. While there is no evident pre-trend leading up to November 2016 in employment, this does not guarantee that the age-specific differences would not show up later in the life of the two cohorts (post-treatment parallel trends), and thus be confused with the implications of the treatment. We address this problem using the triple difference design which accounts for the potential implications of differential effects of age. As we can see in Figures 3c

and 3d differencing out the age effects addresses the pre-trends in earnings. While the pre-trends in the employment event study (Figures 3a and 3b) are not affected, the estimates suggest that the small negative effects which are assigned to the additional employment protection in difference-in-differences estimates, disappear in the triple difference specification.

For the full cohort samples irrespective of the outcome, employment or earnings, we do not see any statistically significant or economically meaningful effects of the forthcoming additional employment protection for the treated cohorts, either for men or for women. Table 1 summarizes these results based on Equation 3. Our preferred triple difference employment estimates suggest statistically insignificant treatment effects of -0.0004 and -0.0022 (i.e.: -0.04 and -0.20 percentage points) for men and women, respectively. Given the standard errors we can rule out negative effects as small as 0.5% and 0.6% for the two groups, respectively. Note that similarity of the effect sizes across genders is partially driven by the differences in the baseline employment probabilities with women having about 10 percentage points higher employment rate. The difference is, on the one hand, due to the fact that women in our samples are five years younger than men, and on the other that those out of the labour market are more likely to be completely outside of the administrative sample. It is worth noting that the higher employment level of these younger cohorts of women compared to the older cohorts of men is consistent with data from Polish Labor Force Survey. For earnings the preferred point estimates imply statistically insignificant reductions of 0.1% for men and statistically insignificant increases of 0.3% for women. Given the standard errors we can rule out negative effects as small as 0.9% and 0.3% for the two groups, respectively. Here, unlike for employment, we find that average earnings of women are about 13% lower than those of men. Overall, we view even the lower bounds of these effects as economically very small.

Importantly, note that the double difference estimates presented in Table 1 suggest negative, larger, and statistically significant at conventional level treatment effects for both employment and earnings.¹⁰ Figure 3 clearly illustrates why this occurs. For earnings we see a negative pre-trend that continues into the post-treatment period, while for employment there is no pre-trend but the post-reform treatment effects are larger compared to the triple difference design. Both facts are consistent with potential negative labor market effects of ageing which is corrected by taking the third difference using adjacent cohorts. It is worth noting though, that even these downward biased estimates are relatively small and imply effect sizes that do not exceed 0.3pp in the case of employment and 1% in the case of earnings.

In the next step, we split the sample on the basis of pre-reform employment status into those

¹⁰The difference in differences estimating equation is $Y_{ict} = \alpha + \theta \text{Post}_{it} \times T_{ic} + \gamma_t + \delta_c + \varepsilon_{ict}$ where Post_{it} takes value of one for September-February 2017-2018 and zero for May-October 2016 while T_{ic} , γ_t , and δ_c are defined as in Equation 1.

with stable employment history, and those out of employment in order to examine the degree of homogeneity of employment results across these samples. We define individuals as in “stable employment” if they are continuously employment between July and September while those who in at least two of the three months were unemployed are considered as not employed. The same conditions are applied for the treatment and the control samples both in the “true reform” and “placebo reform” samples for the triple difference estimation. The triple difference event studies are shown in Figures 3e and 3f for men and women, respectively, and the results are summarized in Table 1. We see small, negative but statistically significant at conventional levels results for the non-employed samples both in the case of men and women, with the estimates of -0.57 and -0.91 percentage points (pp) respectively, and these negative effects clearly evolve over the hold-up period between the time of passing of the retirement legislation in the parliament and the time it took effect in October 2017. They then stabilize in subsequent periods when treatment group enters the employment protection. In contrast, the reform effects for the employees’ samples are modestly positive. As shown in Table 1 employment among men grows by 0.67 pp and among women by 0.19 pp. While we cannot directly distinguish the supply and demand factors behind these results, they are consistent with a positive supply side reaction that outweighs any of the negative demand effects stipulated in the extant EPL literature. Our more detailed heterogeneity analysis, presented in Section 7, confirms this intuition since the positive effects appear to be concentrated among incumbent workers in more precarious jobs. These are the workers who have the most to gain from maintaining their employment up to the point when they reach the age in which they begin to be covered by the additional employment protection.

The results for the full sample, and in particular those for the sample of employees stand in contrast to the theoretical predictions and the two aforementioned studies focusing on older workers. For example, the simulation exercise in Deelen and Jongen (2009) suggests that increasing EPL index from the level of Denmark (1.4) to that of the Netherlands (2.1) could decrease labor force participation of elderly women by as much as 10%. Compared to these even the 95% confidence interval lower bounds in our study are minute. Behaghel et al. (2008) does not provide effects for employment or earnings but his effects on transitions from unemployment to employment for older people are very large at up to -53%. On the other hand, Saez et al. (2023) finds declines in earnings of 8%, conditional on staying on the job, as a result of removing the EPL. This, likewise, is orders of magnitude larger than externalities which we document in this work. The effects identified for the non-employed samples in our analysis are consistent with negative implications of additional protection offered to the eligible groups. While these effects are relatively small, they are by no means irrelevant and it seems they ought to draw attention from the point of view of employment policy including alternative paths to stable employment among older job seekers.

6 Robustness analysis

Our main results presented in Section 5 are based on samples conditional on a number of selection criteria. These include principally conditioning on not being a recipient of an early (“bridge pension”) retirement or a disability pension in the month in which individuals turn 62 (men) or 57 (women). Furthermore, the treatment sample used for our main results consists of ten monthly birth cohorts, which in October 2017 gained the most in terms of additional employment protection out of the 24 and 27 cohorts of men and women respectively who in that month gain extra months of protection as a result of the reform. For clarity we refer to these main samples as A1 (samples used for analysis of employment) and A2 (samples used for analysis of earnings). In the robustness analysis we show results based on samples constructed using alternative criteria (see Online Appendix Tables A.1 and A.2 for details). In particular:

- we limit the samples conditional on receiving either the “bridge” retirement pension or a disability pension at a specific age (main samples A1/A2, and robustness samples: D1/D2 and G1/G2)
- we extend the samples by excluding only those who at a specific age receive the “bridge” pensions (samples B1/B2, E1/E2 and H1/H2);
- we limit the samples further by excluding all individuals who at a specific age receive any social security pension (samples C1/C2, F1/F2 and I1/I2);
- we further narrow down the samples by focusing on the four youngest monthly cohorts among the treated, i.e. those born from July to October (and correspondingly in all other samples for the double and triple difference analysis, samples D1/D2, E1/E2 and F1/F2);
- we condition on the recorded receipt of a specific type of pension not at the age of 62/57 (men/women) as in the main samples (A1/A2) but rather at 61.5/56.5 (men/women, samples G1/G2, H1/H2, I1/I2).

The last two conditions are related to the fact that in our data we can only identify a specific type of pension beginning in 2016. This implies that for the oldest cohorts included in the analyses (men born in 1955 and women born in 1960, i.e. the “placebo reform” treated samples) we can identify their pension receipt when they turned 61.5 and 56.5 only for those born in the second half of the calendar year. There are two important reasons why conditioning on pension receipt at a younger age might play a role. First of all, those who lose their jobs might make an extra effort to secure incomes from pensions, thus making pension receipt endogenous with respect to treatment. Narrowing the age window thus limits this source of potential bias. Secondly, as

we noted in the data section, when we condition on pension receipt at the age of 62/57, in the case of the youngest cohorts (men born in 1959 and women born in 1964, i.e. the “true reform” control samples) we draw on the pension information from as late as October 2020. Lowering the conditioning age by six months means that the latest month of pension data we use for these samples is April 2020, which makes it unlikely to be affected by the COVID-19 pandemic in a major way.

Results of our robustness analyses are presented in Table 2, where we only show regression results using the triple differences design. The first panel shows results using the main sample criteria: including ten monthly birth cohorts and conditioning on pension receipt at the age of 62/57. In the first row we report once again the results from our preferred specification, removing from the sample recipients of either “bridge pensions” or disability pensions. We can see that changing the sample definition by conditioning it on other types of pension makes essentially no difference to the estimates, for employment as well as for earnings. Focusing the analysis on the youngest treated monthly cohorts, those born from July to October, while maintaining the age at which we condition pension receipt at 62/57 makes no difference to the results for men, but translates to larger and statistically significant negative results for employment among women (second panel of Table 2). The estimated effects are in the range of -0.48pp and -0.55pp. Conditioning pension receipt at the lower age (61.5/56.5), while maintaining the narrower number of birth cohorts makes essentially no difference, and here again we find that the implications of the extended employment protection for women are statistically significant and larger in comparison to the results in our main specification.

7 Heterogeneity

In Section 5 we showed that while the overall effects on employment and earnings are statistically insignificant and of economically very low magnitude, the split by pre-reform labor market status shows that these results are a net effect of positive and negative implications of approaching additional employment protection eligibility on, respectively, the employed and the non-employed individuals. Given these findings, for those who were employees prior to the passage of the retirement age legislation in November 2016, we examine the implications of the approaching EPL eligibility among different types of firms, as well as by the level of individual earnings and by local unemployment. We hypothesize that potential positive effects could be most pronounced for lower earnings workers, workers in high turnover firms, workers in firms with relatively young workforce, workers in smaller firms, and workers in high unemployment areas. Our logic behind these hypotheses stems from the fact that marginal increase in effort for those individuals could sway their employers to retain them past the employment protection

cutoff. Conversely, employees in firms that have low turnover, preference for older workers, are large or are located in competitive markets might be less worried about being laid off in the first place.

Our individual level administrative data contains firm-level identifiers which allows us to define such firm-level characteristics as firm size (by employment), firm employment turnover and firm-level age composition of the workforce. In each case we conduct the analysis separately for each tertial of individuals defined on the basis of these firm characteristics specified for firms with at least 10 employees. We measure the characteristics prior to the reform. Additionally, we divide all employees into tertials by their individual level of earnings, and we split the entire sample (employed and non-employed individuals) by the local level of and change in the rate of registered unemployment.¹¹

Results of our heterogeneity analysis (shown in Table 3) seem to confirm our hypothesis of the mechanisms which underline the (net) positive employment protection effects among the sample of employees. In particular, we find that among men the largest positive effects are found for individuals employed in firms with the highest turnover (positive employment effect of 1.28pp in the top tertial) and for those in the lowest tertial of earnings (1.12pp). Although among women we find no heterogeneity in the results by firm characteristics, it appear to be the case that women in the lowest earnings tertial are most positively affected (0.74pp). The results by firm-level turnover and individual earnings tertials are consistent with the hypothesis that employed workers who approach the threshold age of additional EPL eligibility exert extra effort in order to signal their value to the firm and keep their job to ensure that they maintain employment up to the protected age range. This supply reaction on behalf of the employees of course does not exclude the fact that at the same time employers are also laying off others having in mind the approaching limitations concerning their continued employment. Nevertheless, the net effects for the incumbent employees, while relatively low, are positive and question the basic intuition of negative external implications of the additional employment protection.

8 Conclusions

There is continued controversy regarding the implications of regulations which offer different forms of protection for employees vis-à-vis their employers. On the one hand, since labor market protection usually operates through increasing costs of lay-offs, it limits employers' scope for

¹¹We split the counties (powiats) into tertials by unemployment rate (TU1, TU2, TU3) and by change in the rate of unemployment between 2015 and 2016 (TUCH1, TUCH2, TUCH3). We then allocate individuals to unemployment tertials defined as T1: (TU1 and TUCH1; TU1 and TUCH2, TU2 and TUCH1), T2: (TU2 and TUCH2; TU3 and TUCH1, TU1 and TUCH3), and T3: (TU3 and TUCH3; TU3 and TUCH2, TU2 and TUCH3). Defining unemployment tertials differently makes no difference to the results.

adjustment of their employment strategies and production technology. Moreover higher levels of protection might encourage employees to reduce effort and hence lower their productivity. On the other hand, added job security could actually increase worker motivation and output through higher expected pay-offs from their investments in human capital via on- or off-the-job training. Greater job security could also have important social implications if different types of workers have varying degrees of risk aversion and/or mobility. These arguments seem particularly important with regard to older workers who, on the one hand, may be at greater risk of job loss due to deteriorating health and lower productivity, and on the other face difficulties in finding new jobs as a result of being less mobile and because of age discrimination. Policy-wise the debate on age specific employment protection is likely to gain in relevance in the coming years given rapid aging of the population in most developed countries and increasing efforts of many government to extend working lives.

In this paper we use administrative individual level data from Poland to document the external effects of age-specific employment protection legislation for individuals who approach the eligibility threshold. The regulations offer strict protection with regard to employment and earnings and cover employees with less than four years prior to reaching retirement age. Our identification strategy relies on a reform which took effect in October 2017 and reduced retirement age. As a side effect it extended employment protection to younger cohorts of individuals and since it was legislated almost a year earlier, employers - and employees - had plenty of time to adjust their strategies to the approaching change in regulations.

Overall we find no economically meaningful effects of additional protection either with respect to employment or earnings among those who gain additional months of labor market protection. These total effects, however, hide diverging patterns among the subsamples of incumbent employees and those in the relevant age group who are non-employed. For the first group we find small, statistically significant positive effects on employment, and among the latter small negative effects. The positive effects are stronger for men, for whom we find an average effect of about 0.67 percentage point, which is more than three times higher than the average effect for women (0.19pp). Additional heterogeneity analysis we conducted seems to support the hypothesis that the positive effects among incumbent employees derive from increased efforts among those approaching eligibility compared to the control cohorts. We find that the effects are particularly strong among those in more precarious jobs - employees in the lowest tertial of earnings and employed in firms with highest employee turnover. These employees would have most to gain from reaching the age-specific protection threshold and would therefore have extra motivation to remain employed in the months running up to October 2017.

The negative implications with regard to employment among the non-employed suggest that

non-employed women (-0.91pp) suffer more as a result of the regulations in comparison to men (-0.57pp). In fact, women who approach eligibility seem to be at a higher risk of suffering the negative consequences of additional employment protection. In our robustness analysis we show that we can identify negative and statistically significant overall results among the treated cohorts of women when we focus on those who, as a result of the reform, gained the highest number of months of protection (the youngest among our treated cohorts). While the overall effect for these women (born between July-October 1963) is still relatively small - about 0.50 percentage point - it is by no means negligible from the point of view of the overall employment patterns.

Our results seem to suggest that the additional employment protection extended to older workers in Poland, on average, does not have negative employment consequences for those approaching eligibility. On the contrary, those who are most likely to gain from being covered by it seem to exert extra effort to ensure that their employment continues beyond the eligibility threshold. Combined with positive employment effects for the covered workers implied by the additional protection, our results suggest that age-specific employment protection could help in keeping elderly workers attached to the labor market. This finding has three important economic and societal implications. First, employed older workers keep contributing to their pensions rather than relying on welfare or early retirement benefits. This reduces strain on the social security system and in the medium run increases the level of pension payouts that workers could expect post-retirement. Second, there is evidence (see e.g., [\(Atalay et al. 2019\)](#) or [\(Filomena and Picchio 2023\)](#)) that employment could be beneficial for workers nearing retirement when it comes to their cognition and health, which in turn reduces potential burden for the healthcare and social assistance systems. Taken together these two factors could contribute to healthier and wealthier elderly population which relies less on the state. Finally, expanded exposure to productive, older workers could help overcome the previously documented ageism when it comes to demand for labor. On the other hand, the small negative results identified for the non-employed individuals approaching eligibility should be taken into consideration in the design of employment support policies for older workers looking for employment. This may be particularly important in the light of recent changes in the labor code implemented in September 2023, which increased the potential costs for employers of protected workers and in many cases allow employees to return to work already during the litigation process.

References

- Abowd, John and Francis Kramarz (2003) “The Costs of Hiring and Separations,” *Labour Economics*, Vol. 10, No. 5, pp. 499–530.
- Abraham, Katharine and Susan Houseman (2020) “Policies to Improve Workforce Services for Older Americans,” *Economic Studies at Brookings*.
- Albanese, Andrea and Bart Cockx (2019) “Permanent Wage Cost Subsidies for Older Workers. An Effective Tool for Employment Retention and Postponing Early Retirement?,” *Labour Economics*, Vol. 58, pp. 145–166.
- Atalay, Kadir, Garry Barrett, and Anita Staneva (2019) “The Effect of Retirement on Elderly Cognitive Functioning,” *Journal of Health Economics*, Vol. 66, pp. 37–53.
- Autor, David, William Kerr, and Adriana Kugler (2007) “Does Employment Protection Reduce Productivity? Evidence from US States,” *Economic Journal*, Vol. 117, No. 521, pp. F189–F217.
- Bassanini, Andrea and Andrea Garnero (2020) “Employment Protection, Investment, and Firm Growth,” *Review of Financial Studies*, Vol. 33, No. 2, pp. 644–688.
- Bassanini, Andrea, Luca Nuziata, and Danielle Venn (2009) “Job Protection Legislation and Productivity Growth in OECD Countries,” *Economic Policy*, Vol. 24, No. 58, pp. 349–402.
- Behaghel, Luc, Eve Caroli, and Muriel Roger (2014) “Age-biased Technological and Organizational Change, Training and Employment Prospects of Older Workers,” *Economica*, Vol. 81, No. 322, pp. 368–389.
- Behaghel, Luc, Bruno Crépon, and Béatrice Sédillot (2008) “The Perverse Effects of Partial Employment Protection Reform: The Case of French Older Workers,” *Journal of Public Economics*, Vol. 92, No. 3–4, pp. 696–721.
- Bendick, Marc, Lauren Brown, and Kennington Wall (1999) “No Foot in the Door: An Experimental Study of Employment Discrimination Against Older Workers,” *Journal of Aging and Social Policy*, Vol. 10, No. 4, pp. 5–23.
- Bendick, Marc, Charles Jackson, and Horacio Romero (1997) “Employment Discrimination Against Older Workers: An Experimental Study of Hiring Practices,” *Journal of Aging and Social Policy*, Vol. 8, No. 4, pp. 25–46.
- Bertoni, Marco and Giorgio Brunello (2021) “Does a Higher Retirement Age Reduce Youth Employment?,” *Economic Policy*, Vol. 36, No. 106, pp. 325–372.
- Bjuggren, Carl (2018) “Employment Protection and Labor Productivity,” *Journal of Public Economics*, Vol. 157, pp. 138–157.
- Boeri, Tito and Pietro Garibaldi (2007) “Two Tier Reforms of Employment Protection: A Honey-moon Effect?,” *Economic Journal*, Vol. 117, No. 521, pp. F357–F385.

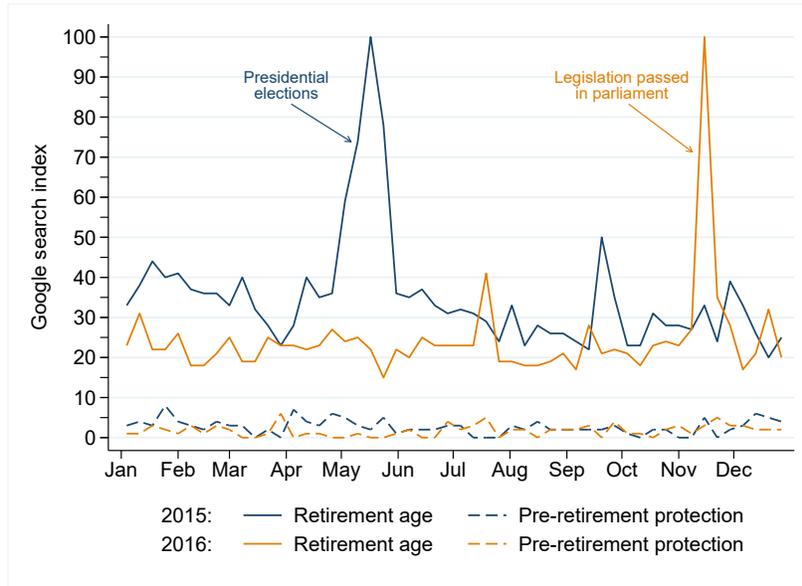
- Boockmann, Bernhard, Thomas Zwick, Andreas Ammermüller, and Michael Maier (2012) “Do Hiring Subsidies Reduce Unemployment Among Older Workers? Evidence from Natural Experiments,” *Journal of the European Economic Association*, Vol. 10, No. 4, pp. 735–764.
- Börsch-Supan, Axel and Matthias Weiss (2016) “Productivity and Age: Evidence from Work Teams at the Assembly Line,” *Journal of the Economics of Ageing*, Vol. 7, pp. 30–42.
- Brown, Jeffrey, James Poterba, and David Richardson (2022) “Trends in Retirement and Retirement Income Choices by TIAA Participants: 2000-2018,” NBER WP 29946.
- Cahuc, Pierre, Franck Malherbet, and Julien Prat (2019) “The Detrimental Effect of Job Protection on Employment: Evidence from France,” IZA DP 12384.
- Daniel, Kirsten and John Heywood (2007) “The Determinants of Hiring Older Workers: UK Evidence,” *Labour Economics*, Vol. 14, No. 1, pp. 35–51.
- Deaton, Angus and Christina Paxson (1998) “Aging and Inequality in Income and Health,” *American Economic Review*, Vol. 88, No. 2, pp. 248–253.
- Deelen and E Jongen (2009) “Employment Protection, Rethinking Retirement - From Participation towards Allocation,” Technical report, CPB Special Publication, No 80.
- Diebold, Francis, David Neumark, and Daniel Polsky (1997) “Job Stability in the United States,” *Journal of Labor Economics*, Vol. 15, No. 2, pp. 206–233.
- Duhon, Madeline, Edward Miguel, Amos Njuguna, Daniela Pinto Veizaga, and Michael Walker (2023) “Preparing for an Aging Africa: Data-Driven Priorities for Economic Research and Policy,” NBER WP 31750.
- Duval, Romain, Davide Fureri, and Joao Jalles (2020) “Job Protection Deregulation in Good and Bad Times,” *Oxford Economic Papers*, Vol. 72, No. 2, pp. 370–390.
- Eyster, Lauren, Richard Johnson, and Eric Toder (2008) “Current Strategies to Employ and Retain Older Workers,” Technical report, Urban Institute.
- Faberman, Jason and Marianna Kudlyak (2019) “The Intensity of Job Search and Search Duration,” *American Economic Journal: Macroeconomics*, Vol. 11, No. 3, pp. 327–357.
- Filomena, Mattia and Matteo Picchio (2023) “Retirement and Health Outcomes in a Meta-Analytical Framework,” *Journal of Economic Surveys*, Vol. 37, No. 4, pp. 1120–1155.
- Frimmel, Wolfgang, Thomas Horvath, Mario Schnalzenberger, and Rudolf Winter-Ebmer (2018) “Seniority Wages and the Role of Firms in Retirement,” *Journal of Public Economics*, Vol. 164, pp. 19–32.
- Gordo, Laura and Vegard Skirbekk (2013) “Skill Demand and the Comparative Advantage of Age: Job Tasks and Earnings from the 1980s to the 2000s in Germany,” *Labour Economics*, Vol. 22, pp. 61–69.

- Griffith, Rachel and Gareth Macartney (2014) “Employment Protection Legislation, Multinational Firms, and Innovation,” *Review of Economics and Statistics*, Vol. 96, No. 1, pp. 135–150.
- Hairault, Jean-Olivier, Francois Langot, and Thepthida Sopraseuth (2010) “Distance to Retirement and Older Workers’ Employment: The Case for Delaying the Retirement Age,” *Journal of the European Economic Association*, Vol. 8, No. 5, pp. 1034–1076.
- Hakola, Tuulia and Roope Uusitalo (2005) “Not so Voluntary Retirement Decisions? Evidence from a Pension Reform,” *Journal of Public Economics*, Vol. 89, No. 11-12, pp. 2121–2136.
- Hardy, Wojciech, Anna Kiełczewska, Piotr Lewandowski, and Iga Magda (2018) “Job Retention Among Older Workers in Central and Eastern Europe,” *Baltic Journal of Economics*, Vol. 18, No. 2, pp. 69–94.
- Hijzen, Alexander, Leopoldo Mondauto, and Stefano Scarpetta (2017) “The Impact of Employment Protection on Temporary Employment: Evidence from a Regression Discontinuity Design,” *Labour Economics*, Vol. 46, pp. 64–76.
- Huttunen, Kristiina, Jukka Pirttilä, and Roope Uusitalo (2013) “The Employment Effects of Low-wage Subsidies,” *Journal of Public Economics*, Vol. 97, pp. 49–60.
- Ichino, Andrea and Regina Riphahn (2005) “The Effect of Employment Protection on Worker Effort: Absenteeism During and After Probation,” *Journal of the European Economic Association*, Vol. 3, No. 1, pp. 120–143.
- Ilmakunnas, Pekka and Seija Ilmakunnas (2015) “Hiring Older Employees: Do the Age Limits of Early Retirement and the Contribution Rates of Firms Matter?,” *Scandinavian Journal of Economics*, Vol. 117, No. 1, pp. 164–194.
- Jacob, Brian (2013) “The Effect of Employment Protection on Teacher Effort,” *Journal of Labor Economics*, Vol. 31, No. 4, pp. 727–761.
- Jimeno, Juan, Juan Rojas, and Sergio Puente (2008) “Modelling the Impact of Aging on Social Security Expenditures,” *Economic Modeling*, Vol. 25, No. 2, pp. 201–224.
- Kahn, Lawrence (2007) “The Impact of Employment Protection Mandates on Demographic Temporary Employment Patterns: International Microeconomic Evidence,” *Economic Journal*, Vol. 117, No. 521, pp. F333–F356.
- (2010) “Employment Protection Reforms, Employment and the Incidence of Temporary Jobs in Europe: 1996-2001,” *Labour Economics*, Vol. 17, pp. 1–15.
- Kan, Kamhon and Yen-Ling Lin (2011) “The Effects of Employment Protection on Labor Turnover: Empirical Evidence from Taiwan,” *Economic Inquiry*, Vol. 49, No. 2, pp. 398–433.
- Kitao, Sagiri (2014) “Sustainable Social Security: Four Options,” *Review of Economic Dynamics*, Vol. 17, No. 4, pp. 756–779.
- Kotschy, Rainer and David Bloom (2023) “Population Aging and Economic Growth: From Demographic Dividend to Demographic Drag?,” NBER WP 31585.

- Krueger, Alan and Jörn-Steffen Pischke (1992) “The Effect of Social Security on Labor Supply: A Cohort Analysis of the Notch Generation,” *Journal of Labor Economics*, Vol. 10, No. 4, pp. 412–437.
- Kugler, Adriana and Giovanni Pica (2008) “Effects of Employment Protection on Worker and Job Flows: Evidence from the 1990 Italian Reform,” *Labour Economics*, Vol. 15, pp. 78–95.
- Lahey, Joanna (2008) “Age, Women, and Hiring: An Experimental Study,” *Journal of Human Resources*, Vol. 43, No. 1, pp. 30–56.
- Laun, Lisa and Marten Palme (2023) “Pension Reform, Incentives to Retire and Retirement Behavior: Empirical Evidence From Swedish Micro-Data,” NBER WP 31800.
- Lazear, Edward (1990) “Job Security Provisions and Employment,” *Quarterly Journal of Economics*, Vol. 105, No. 3, pp. 699–726.
- Leonardi, Marco and Giovanni Pica (2013) “Who Pays for It? The Heterogeneous Wage Effects of Employment Protection Legislation,” *Economic Journal*, Vol. 123, No. 573, pp. 1236–1278.
- Mahlberg, Bernhard, Inga Freund, Jesus Cuaresma, and Alexia Prskawetz (2013) “Ageing, Productivity and Wages in Austria,” *Labour Economics*, Vol. 22, pp. 5–15.
- Martins, Pedro (2009) “Dismissals for Cause: The Difference That Just Eight Paragraphs Can Make,” *Journal of Labor Economics*, Vol. 27, No. 2, pp. 257–279.
- Messina, Julian and Giovanna Vallanti (2007) “Job Flow Dynamics and Firing Restrictions: Evidence from Europe,” *Economic Journal*, Vol. 117, No. 521, pp. F279–F301.
- Morris, Todd and Benoit Dostie (2023) “Graying and Staying on the Job: The Welfare Implications of Employment Protection for Older Workers,” IZA DP 16430.
- Neumark, David, Ian Burn, and Patrick Button (2019) “Is It Harder for Older Workers to Find Jobs? New and Improved Evidence from a Field Experiment,” *Journal of Political Economy*, Vol. 127, No. 2, pp. 922–970.
- OECD (2019) “Working Better with Age,” Technical report, OECD, Paris.
- Perek-Białas, Jolanta and Konrad Turek (2012) “Organisation-level Policy Towards Older Workers in Poland,” *International Journal of Social Welfare*, Vol. 21, No. S1, pp. S101–S116.
- Rabaté, Simon (2019) “Can I Stay or Should I Go? Mandatory Retirement and the Labor-force Participation of Older Workers,” *Journal of Public Economics*, Vol. 180, p. 104078.
- Saez, Emmanuel, Benjamin Schoefer, and David Seim (2023) “Deadwood Labor: The Effect of Eliminating Employment Protection,” NBER WP 31797.
- Sestito, Paolo and Eliana Viviano (2018) “Firing Costs and Firm Hiring: Evidence from an Italian Reform,” *Economic Policy*, Vol. 33, No. 93, pp. 101–130.

- Staubli, Stefan and Josef Zwimüller (2013) “Does Raising the Early Retirement Age Increase Employment of Older Workers,” *Journal of Public Economics*, Vol. 108, pp. 17–32.
- van der Wiel, Karen (2010) “Better Protected, Better Paid: Evidence on How Employment Protection Affects Wages,” *Labour Economics*, Vol. 17, pp. 16–26.
- Yoo, Gyeongjoon and Changhui Kang (2012) “The Effect of Protection of Temporary Workers on Employment Levels: Evidence from the 2007 Reform of South Korea,” *ILR Review*, Vol. 65, No. 3, pp. 578–606.

Figure 1: Social reaction to retirement age reform initiatives: "retirement age" and "pre-retirement protection" in online searches in 2015 and 2016

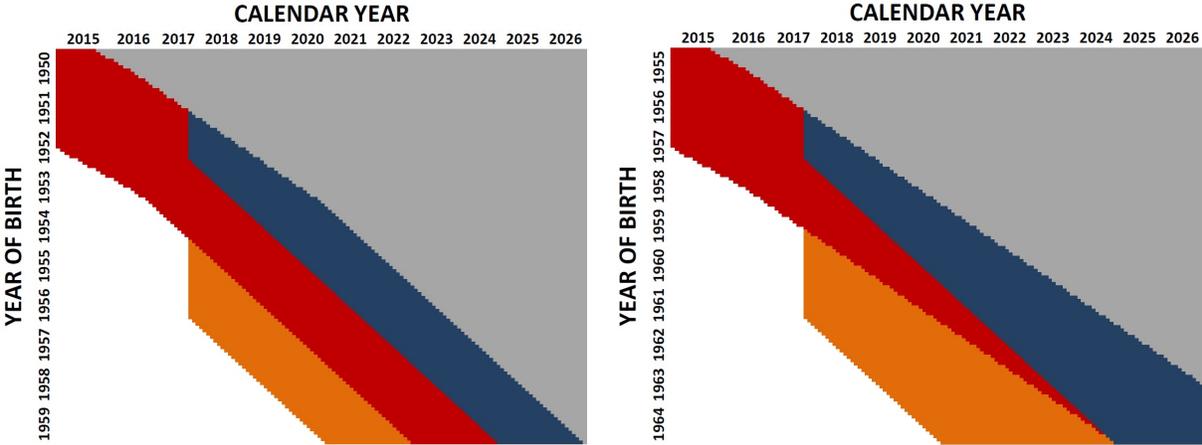


Note: Google search statistics for "wiek emerytalny" (EN: "retirement age") and "ochrona przedemerytalna" (EN: "pre-retirement protection") based on Google Trends.

The two rounds of the presidential election took place in Poland on May 10th and 24th 2015 with the pledge of the opposition candidate, Andrzej Duda, to return to lower retirement age of 60 and 65 for women and men, respectively. Parliamentary elections followed on October 25th 2015 and were won by the Law and Justice (PiS) party who supported Andrzej Duda in the presidential vote. The first reading of the presidential legislative initiative took place December 2015, the second reading nearly a year later on November 15th 2016, with the legislation passed on the following day. See main text for more details.

Source: Google Trends.

Figure 2: Retirement age and employment protection by cohort and time



(a) Men

(b) Women

Note: The grey area reflects cohort-specific retirement age eligibility to public pensions in accordance with the 2012 reform which gradually increased retirement age from 60/65 (women/men) to 67. The corresponding red area reflects the months of eligibility to age-related employment protection. The navy-blue area (overlapping partly with the red area) shows the additional earlier eligibility to pension reforms following the 2016 reform which reduced retirement age back to 60/65. The orange area is the additional age-related EPL eligibility gained as a result of the retirement age reform which came into effect in October 2017.

Source: Authors' illustration on the basis of the corresponding legislation.

Table 1: Pre-retirement employment protection and effects for employment and earnings among cohorts approaching eligibility: Average treatment effects

	Men	Women
Employment:		
Full sample mean#:	0.6656	0.7641
Full sample estimates, DID:	-0.0033*** (0.0011)	-0.0027*** (0.0010)
Full sample estimates, DIDID:	-0.0004 (0.0015)	-0.0022 (0.0014)
Employees sample estimates, DIDID	0.0067*** (0.0014)	0.0019* (0.0011)
Non-employed sample estimates, DIDID	-0.0057* (0.0034)	-0.0091** (0.0043)
Earnings (logs):		
Earnings sample mean#:	8.071	7.939
Earnings sample estimates, DID:	-0.0099*** (0.0029)	-0.0081*** (0.0024)
Earnings sample estimates, DIDID	-0.0010 (0.0041)	0.0030 (0.0033)

Notes: DID - difference-in-differences estimates, DIDID - triple difference estimates (in accordance with Equation 3). Samples of employees and non-employed defined on the basis of labor market status in the months July-September prior to treatment (and in corresponding months for respective samples in the triple difference estimation). Based on samples A1 (employment) and A2 (earnings), see main text for definition of samples and Tables A.1 and A.2 for details. # - mean values calculated for the treated cohorts for months considered as pre-treatment (May-October 2016).

Source: Authors' calculations using Ministry of Finance administrative database.

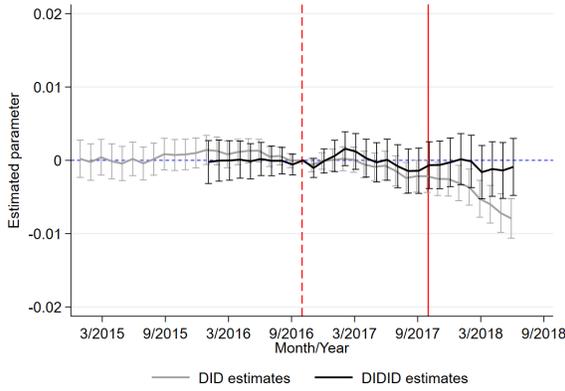
Table 2: Pre-retirement employment protection and effects for employment and earnings among cohorts approaching eligibility: Robustness analysis

DIDID estimates	Employment:		Earnings:	
	Men	Women	Men	Women
Main sample criteria (months 1-10)				
Sample 62/57, samples A1/A2	-0.0004 (0.0015)	-0.0022 (0.0014)	-0.0010 (0.0041)	0.0030 (0.0033)
Sample 62/57, samples B1/B2	0.0001 (0.0018)	-0.0014 (0.0015)	0.0016 (0.0042)	0.0045 (0.0034)
Sample 62/57, samples C1/C2	-0.0009 (0.0015)	-0.0019 (0.0014)	-0.0037 (0.0040)	0.0030 (0.0033)
Robustness estimates: (months 7-10)				
Sample 62/57, samples D1/D2	0.0003 (0.0025)	-0.0055** -0.0023	-0.0064 (0.0066)	0.0051 -0.0052
Sample 62/57, samples E1/E2	0.0010 (0.0028)	-0.0051** -0.0025	-0.0043 (0.0068)	0.0079 -0.0054
Sample 62/57, samples F1/F2	-0.0013 (0.0024)	-0.0048** -0.0023	-0.0064 (0.0066)	0.0001 -0.0052
Robustness estimates: (months 7-10)				
Sample 61.5/56.5, samples G1/G2	0.0006 (0.0025)	-0.0051** (0.0023)	-0.0061 (0.0066)	0.0045 (0.0052)
Sample 61.5/56.5, samples H1/H2	0.0012 (0.0028)	-0.0050** (0.0025)	-0.0040 (0.0068)	0.0077 (0.0054)
Sample 61.5/56.5, samples I1/I2	-0.0015 (0.0024)	-0.0048** (0.0023)	-0.0067 (0.0066)	0.0001 (0.0052)

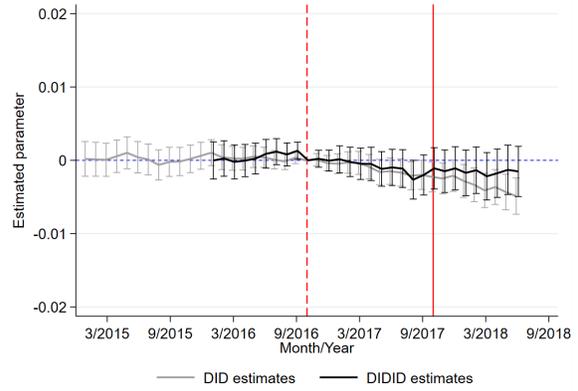
Notes: All values show triple difference estimates (DIDID) of the average treatment effect as specified in Equation 3. First row shows the main specification results from Table 1 for comparison. See main text for the definitions of the analysis samples and Tables A.1 and A.2 for details of sample sizes.

Source: Authors' calculations using Ministry of Finance administrative database.

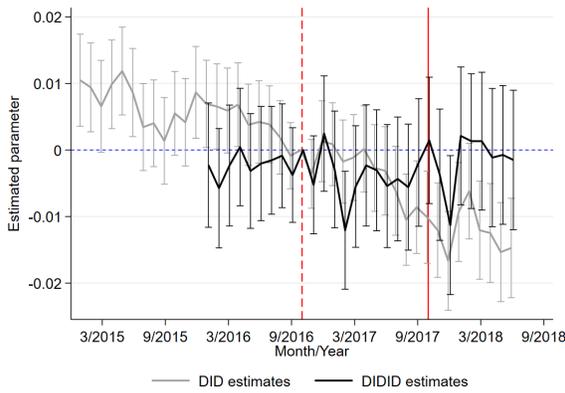
Figure 3: Pre-retirement employment protection and effects for employment and earnings among cohorts approaching eligibility: Event studies



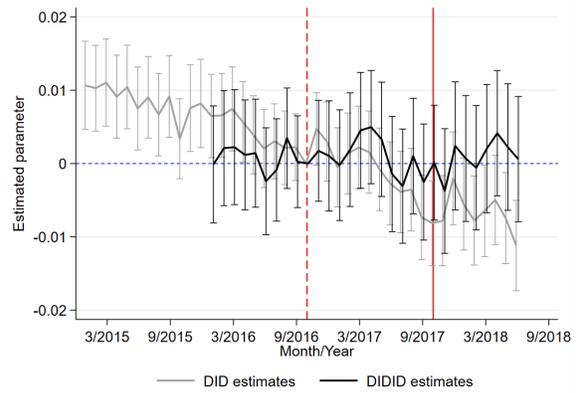
(a) Employment: Men



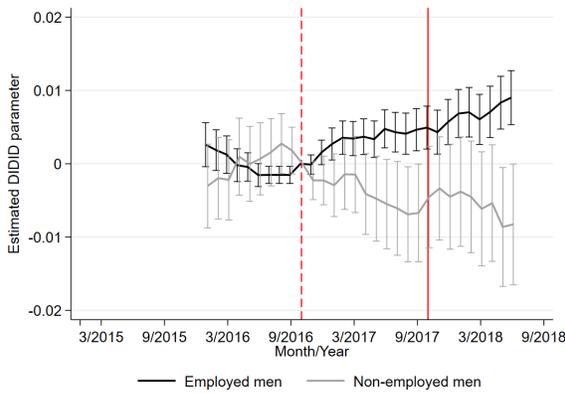
(b) Employment: Women



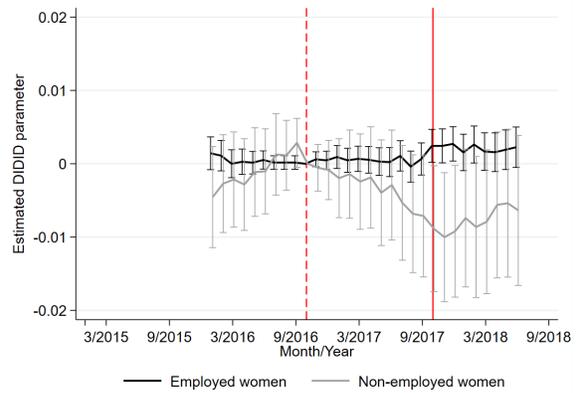
(c) Labor earnings (logs): Men



(d) Labor earnings (logs): Women



(e) Employment, men: employees & non-employed



(f) Employment, women: employees & non-employed

Notes: DID - difference-in-differences estimates; DIDID - triple difference estimates. Figures shifted marginally to avoid overlap for better visual clarity. The dashed vertical lines mark October 2016 (last month before legislation of reduced retirement age), while the solid vertical lines mark October 2017 (when legislation becomes binding). Samples of employees and non-employed defined on the basis of labor market status in the months July-September prior to treatment (and in corresponding months for respective samples in the triple difference estimation). See text for details.

Source: Authors' calculations using Ministry of Finance administrative database.

Table 3: Pre-retirement employment protection and effects for employment and earnings among cohorts approaching eligibility: Heterogeneity analysis

DIDID estimates	Men:				Women:			
	Full (sub-) sample	Tertial 1	Tertial 2	Tertial 3	Full (sub-) sample	Tertial 1	Tertial 2	Tertial 3
Employment:								
By firm size:		0.0062** (0.0025)	0.0085*** (0.0022)	0.0038* (0.0020)		0.0029 (0.0019)	-0.0014 (0.0016)	0.0016 (0.0016)
By firm turnover:	0.0062*** (0.0013)	0.0011 (0.0014)	0.0032 (0.0020)	0.0128*** (0.0030)	0.0010 (0.0010)	0.0006 (0.0011)	0.0008 (0.0015)	0.0016 (0.0023)
By firm age composition:		0.0082*** (0.0025)	0.0034* (0.0020)	0.0067*** (0.0022)		0.0019 (0.0020)	0.0014 (0.0015)	-0.0003 (0.0017)
By earnings:	0.0067*** (0.0014)	0.0112*** (0.0033)	0.0050*** (0.0018)	0.0013 (0.0018)	0.0019* (0.0011)	0.0074*** (0.0026)	0.0001 (0.0014)	-0.0023* (0.0012)
By local unemployment:	-0.0004 (0.0015)	-0.0002 (0.0021)	0.0041 (0.0033)	-0.0009 (0.0032)	-0.0022 (0.0014)	-0.0026 (0.0020)	-0.0018 (0.0030)	-0.0016 (0.0028)
Earnings:								
By firm size:		-0.0091 (0.0058)	-0.0044 (0.0069)	-0.0034 (0.0079)		0.0011 (0.0049)	-0.0007 (0.0054)	0.0055 (0.0065)
By firm turnover:	-0.0058 (0.0040)	-0.0101 (0.0066)	0.0019 (0.0065)	-0.0026 (0.0074)	0.0018 (0.0033)	-0.0001 (0.0053)	-0.0004 (0.0054)	0.0063 (0.0061)
By firm age composition:		-0.0105 (0.0081)	0.0034 (0.0066)	-0.0107* (0.0061)		-0.0034 (0.0067)	0.0069 (0.0053)	-0.0020 (0.0048)
By earnings:	-0.0053 (0.0037)	0.0050 (0.0067)	-0.0022 (0.0052)	0.0007 (0.0062)	-0.0008 (0.0031)	-0.0089 (0.0054)	0.0017 (0.0047)	0.0075 (0.0054)
By local unemployment:	-0.0010 (0.0041)	-0.0014 (0.0057)	-0.0110 (0.0085)	0.0051 (0.0078)	0.0030 (0.0033)	-0.0001 (0.0046)	0.0010 (0.0073)	0.0120* (0.0064)

Notes: All values show triple difference estimates (DIDID) of the average treatment effect as specified in Equation 3. See main text for the definitions of the analysis samples and Tables A.1 and A.2 for details.

Source: Authors' calculations using Ministry of Finance administrative database.

Appendix:

Appendix Tables

Table A.1: Sample selection and analysis samples, men
Main, heterogeneity and robustness samples, numbers of individuals

Birth months:	Reform sample:		Placebo sample:		Total sample:
	Treatment 01-10.1956	Control 01-10.1958	Treatment 01-10.1955	Control 01-10.1957	
Full sample (individuals):	155,087	163,536	149,688	160,755	629,066
Sample selection for main specifications:					
- sample after step 1: disability/early retirement at age 62	114,803	128,463	120,262	121,455	484,983
- sample after step 2: sectors of employment (employment estimation sample, A1)	54,397	63,475	56,166	59,031	233,069
Heterogeneity analysis:					
Estimation samples, employment:					
- terciles by firm size, turnover and age composition	29,393	35,163	28,170	31,924	124,650
- terciles by individual earnings (employees)	34,682	41,572	33,365	37,906	147,525
- non-employees	18,787	20,851	21,845	20,217	81,700
Estimation samples, earnings:					
- earnings estimation sample (sample A2)	38,051	45,491	37,055	41,280	161,877
- terciles by firm size, turnover & age composition	29,393	35,163	28,170	31,924	124,650
- terciles by individual earnings	34,682	41,572	33,365	37,906	147,525
Alternative samples for robustness analysis adjusted sample selection criteria in step 1:					
Birth months:	01-10.1956	01-10.1958	01-10.1955	01-10.1957	
Estimation sample, employment:					
B1, step 1: only early retirement pension at age 62	61,626	72,001	59,578	66,972	260,177
C1, step 1: any pension at age 62	49,689	60,837	43,740	55,658	209,924
Estimation sample, earnings:					
B2, step 1: only early retirement pension at age 62	44,595	53,616	41,533	49,152	188,896
C2, step 1: any pension at age 62	36,284	45,289	31,245	41,268	154,086
Birth months:	07-10.1956	07-10.1958	07-10.1955	07-10.1957	
Estimation sample, employment:					
D1, step 1: disability/early retirement at age 62	20,999	23,624	21,617	22,474	88,714
E1, step 1: only early retirement at age 62	23,790	26,846	23,035	25,588	99,259
F1, step 1: any pension at age 62	19,367	22,613	17,140	21,282	80,402
G1, step 1: disability/early retirement at age 61.5	21,400	23,954	22,180	22,830	90,364
H1, step 1: only early retirement at age 61.5	24,034	26,993	23,353	25,753	100,133
I1, step 1: any pension at age 61.5	19,374	22,932	17,140	21,616	81,062
Estimation sample, earnings:					
D2, step 1: disability/early retirement at age 62	14,716	16,986	14,274	15,805	61,781
E2, step 1: only early retirement at age 62	17,238	20,054	16,052	18,878	72,222
F2, step 1: any pension at age 62	14,150	16,874	12,246	15,839	59,109
G2, step 1: disability/early retirement at age 61.5	15,045	17,275	14,708	16,095	63,123
H2, step 1: only early retirement at age 61.5	17,441	20,184	16,301	19,015	72,941
I2, step 1: any pension at age 61.5	14,154	17,157	12,246	16,119	59,676

Notes: See notes in text presented below Table A.2.

Source: Authors' calculations using Ministry of Finance administrative database.

Table A.2: Sample selection and analysis samples, women
Main, heterogeneity and robustness samples, numbers of individuals

Birth months:	Reform sample:		Placebo sample:		Total sample:
	Treatment 01-10.1961	Control 01-10.1963	Treatment 01-10.1960	Control 01-10.1962	
Full sample (individuals):	150,368	148,902	157,283	147,534	604,087
Sample selection for main specifications:					
- sample after step 1: disability/early retirement at age 57	131,632	131,630	147,639	129,261	540,162
- sample after step 2: sectors of employment (employment estimation sample, A1)	60,374	60,329	65,569	60,189	246,461
Heterogeneity analysis:					
Estimation samples, employment:					
- terciles by firm size, turnover and age composition	38,655	38,651	40,644	38,438	156,388
- terciles by individual earnings (employees)	44,576	44,617	46,961	44,499	180,653
- non-employees	14,802	14,772	17,522	14,726	61,822
Estimation samples, earnings:					
- earnings estimation sample (sample A2)	48,004	48,102	51,057	48,136	195,299
- terciles by firm size, turnover & age composition	38,655	38,651	40,433	38,300	156,039
- terciles by individual earnings	44,576	44,617	46,618	44,250	180,061
Alternative samples for robustness analysis:					
adjusted sample selection criteria in step 1					
Birth months:	01-10.1961	01-10.1963	01-10.1960	01-10.1962	
Estimation sample, employment:					
B1, step 1: only early retirement pension at age 57	62,872	63,239	67,067	62,979	256,157
C1, step 1: any pension at age 57	56,814	58,107	57,000	57,528	229,449
Estimation sample, earnings:					
B2, step 1: only early retirement pension at age 57	50,913	51,662	53,517	51,359	207,451
C2, step 1: any pension at age 57	46,646	47,704	46,728	47,471	188,549
Birth months:	07-10.1961	07-10.1963	07-10.1960	07-10.1962	
Estimation sample, employment:					
D1, step 1: disability/early retirement at age 57	22,883	22,476	24,219	22,573	92,151
E1, step 1: only early retirement pension at age 57	23,847	23,585	24,833	23,627	95,892
F1, step 1: any pension at age 57	21,590	21,660	21,264	21,643	86,157
G1, step 1: disability/early retirement at age 56.5	23,038	22,630	24,445	22,736	92,849
H1, step 1: only early retirement pension at age 56.5	23,933	23,655	24,960	23,710	96,258
I1, step 1: any pension at age 56.5	21,595	21,810	21,264	21,800	86,469
Estimation sample, earnings:					
D2, step 1: disability/early retirement at age 57	18,185	17,968	18,933	18,107	73,193
E2, step 1: only early retirement pension at age 57	19,319	19,332	19,870	19,338	77,859
F2, step 1: any pension at age 57	17,720	17,819	17,455	17,920	70,914
G2, step 1: disability/early retirement at age 56.5	18,318	18,100	19,076	18,244	73,738
H2, step 1: only early retirement pension at age 56.5	19,393	19,393	19,975	19,409	78,170
I2, step 1: any pension at age 56.5	17,723	17,952	17,455	18,055	71,185

Notes: See notes in text presented below.

Source: Authors' calculations using Ministry of Finance administrative database.

Notes for Tables A.1 and A.2:

Sample sizes presented in the Tables reflect the number of individuals. ‘Reform sample’ – sample subjected to the retirement age reform in 2016-2017 in the analysis time frame; ‘Placebo sample’ – sample not subjected to the retirement age reform in 2016-2017 in the analysis time frame, used for ‘placebo reform’ in years 2015-2016 for triple difference specification; ‘Full sample’ – individuals born in specific months, registered as being alive at the end of 2018.

- Sample selection, step 1: excluding individuals who at a specific age (62/57 in main sample for men/women, 61.5/56.5 in robustness analysis for men/women) are registered as claiming a specified type of social security pension.

- Sample selection, step 2: excluding individuals who at any point between 01-2015 and 06-2018 are recorded as being employed in selected sectors (see # below) and/or worked in occupations that have a separate pension scheme (uniformed services, judges and public prosecutors).

- Samples for heterogeneity analysis by firm characteristics additionally conditional on being employed in firms with more than 10 employees in the first half of 2016 (“reform sample”) and first half of 2015 (“placebo sample”); firm-level turnover defined on the basis of turnover between first half of 2015 and 2016; age composition of employment defined by the proportion of people age 50+;

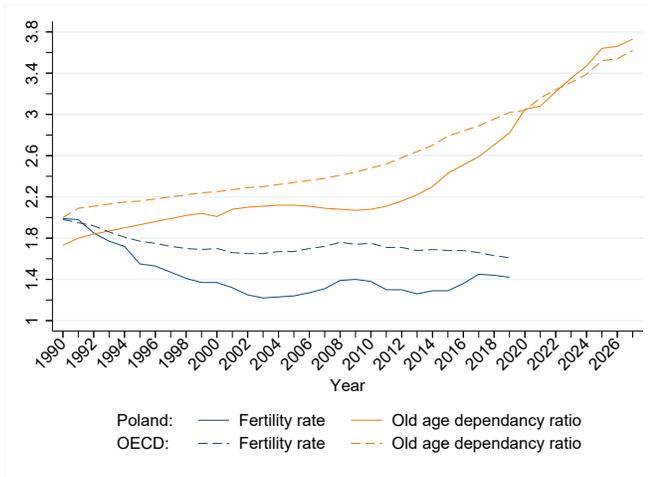
- Samples for analysis of earnings: individuals need to have at least one observation of positive earnings from an employment contract for the estimation time frame;

- Samples for earnings tertials: based on average earnings (from an employment contract) for the months January-June 2016 (“reform sample”) or January-June 2015 (“placebo sample”);

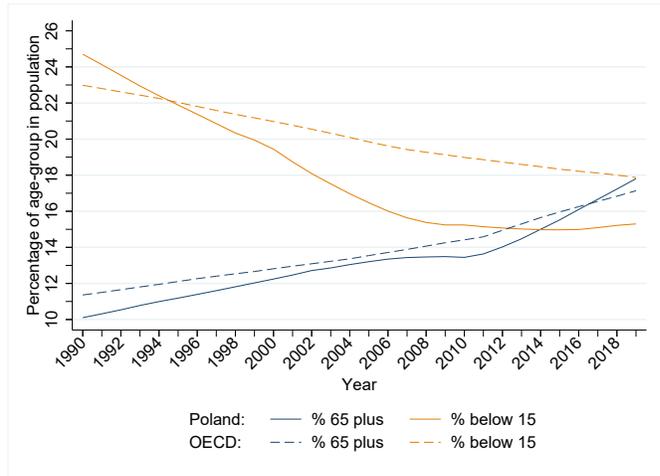
- Employees recorded as working in the following sectors are excluded from the analysis: agriculture, forestry, hunting and fishing (A); mining and quarrying (B); water supply, sewage and waste management and reclamation activities (E); professional, scientific and technical activities (M); public administration and national defence, compulsory social security (O); education (P); households with employees, households producing goods and providing services for their own needs (T); extraterritorial organizations and teams (U).

Appendix Figures

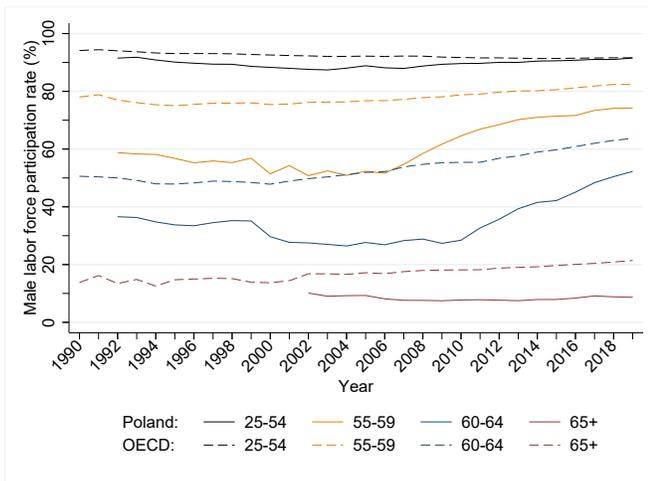
Figure A.1: Demographic trends and labor force participation: Poland and OECD



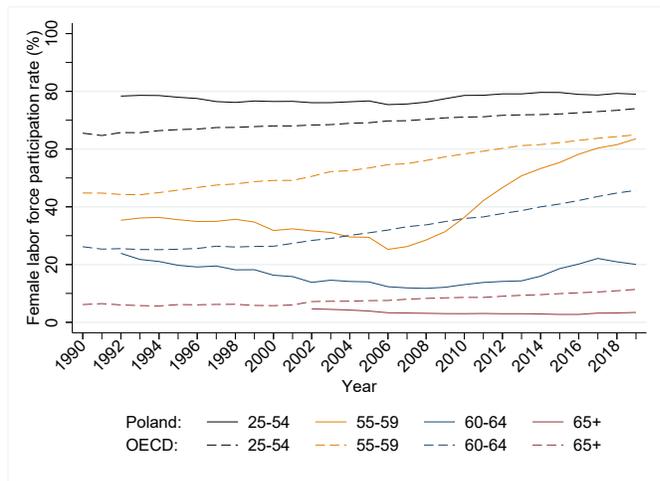
(a) Fertility rates and old age dependency ratios



(b) Percentage of individuals aged below 15 and over 65 years old in total population



(c) Male labor force participation rates



(d) Female labor force participation rates

Note: OECD values represent the average for all OECD countries.
 Source: OECD database.