

# Assessing the role of ageing feminizing & better-educated workforces on TFP growth



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Main objective= assess impact of:

- Ageing
- Feminisation
- Rising educational attainment

on TFP growth rate at the level of the firm

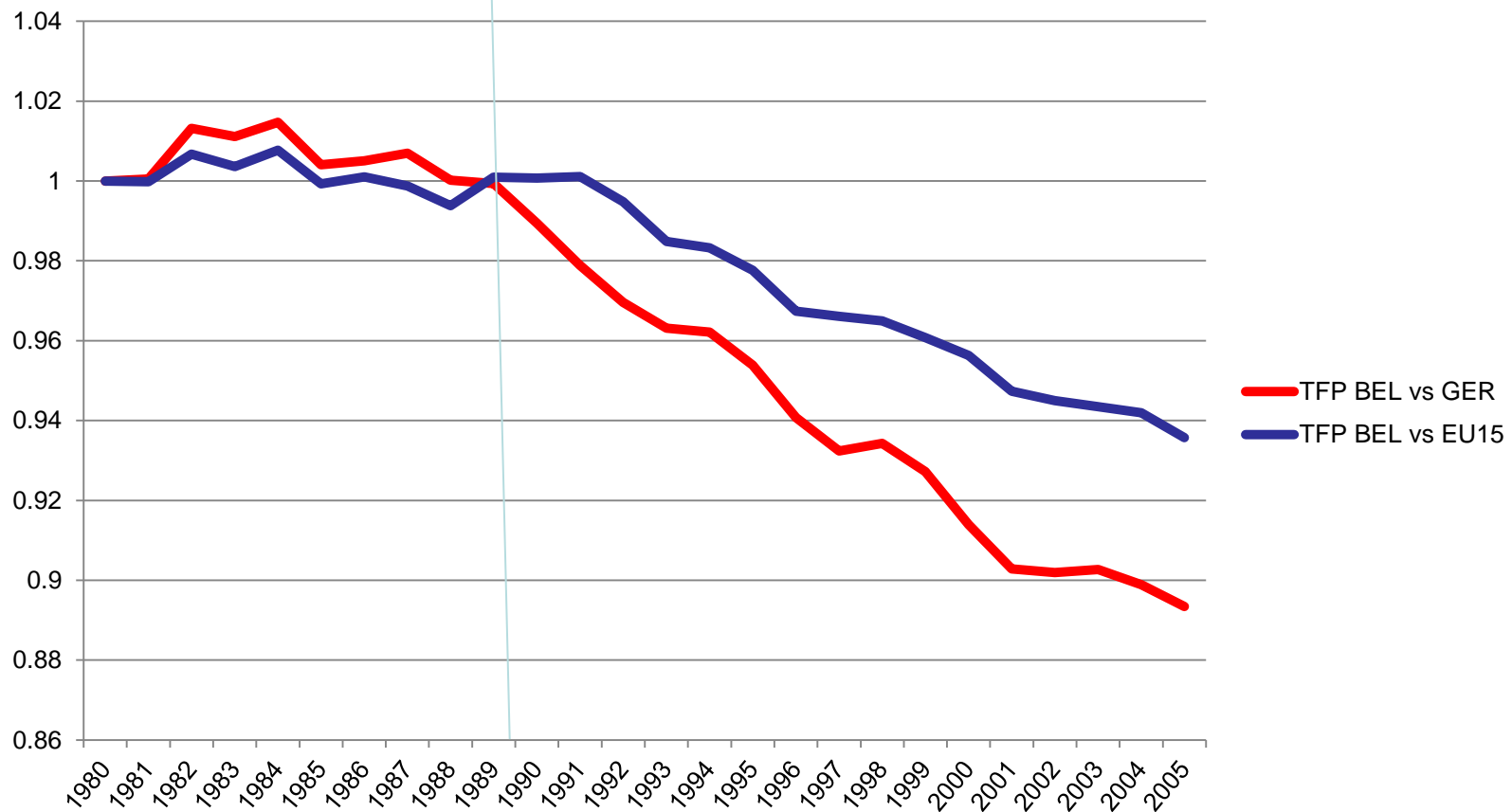
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**Secondary objective** quantify/simulate the likely impact of socio-demographic changes on Belgium's overall TPF growth, retrospectively (1990-today) and prospectively (today-2040) using demographic & employment data/projections

# Motivation

# Belgium has experienced a decrease in TFP growth (Biatour et al., 2011)

Belgian TFP growth with respect to Germany and EU15, 1980-2005



## **Age**

- the share of workers aged 50 and more increased from 18% in 1980 to 21% in 2005
- the share of workers with less than 35 years decreased from more than 30% in 1980 to about 20% in 2005 (Statbel)

## **Gender**

- The share of older women almost doubled passing from more than 3% in 1980 to 6% in 2005 (EU-KLEMS)

## **Education**

- the share of 2-year-college-educated workers has increased from 17.9% to 19.2% (2002-2011)
- university-educated employees from 7.4% to 8% (Vandenberghé & Lebedinski, 2013)

# Contribution

- **Firm-level perspective on TFP growth**
- **Negative effect of age** (BELGIUM/ Lallemand & Rycx, 2009; Cataldi, A., S. Kampelmann & F. Rycx, 2011; Vandenberghe, 2013 ; Vandenberghe, Rigo & Waltenberg, 2012, Vandenberghe, 2011a,b; FRANCE/ Crépon, Deniau & Pérez-Duarte, 2002; USA/ Hellerstein & Neumark, 2007)
- **Gender and Education dimensions of ageing**
- **Focus on TFP rather than on labour productivity or wages**

# Analytical framework



Consider a labour-augmented Cobb-Douglas technology

$$Y_{it} = A_{it} K_{it}^{\alpha} (QL_{it})^{\beta} \quad [1]$$

with  $Q_{it}$  a labour-quality index à la Hellerstein – Neumark (HN), specified as a CES

$$QL_{it} = [\mu_1 (L_{it}^1)^{\rho} + \dots + \mu_n (L_{it}^n)^{\rho}]^{1/\rho} \quad [2]$$

- $L_{itj}$ ,  $j=1 \dots n$  labour types (e.g age, gender, blue-vs white collar categories)
- $\mu_j$  reflects the (relative) marginal productivity of type  $j$  labour
- $\rho$  the CES substitutability parameter
- $A_{it} = A_{i0} e^{\tau \cdot t + \omega_{it}}$ ; with  $A_{i0}$  the starting value of firm  $i$ 's TFP,  $e^{\tau \cdot t + \omega_{it}}$  capturing its dynamic
  - $\tau$  is the common annual rate of growth;
  - $\omega_{it}$  the firm-specific term, with  $\omega_{it} = \Theta_i + \delta_{it}$  containing a fixed effect<sup>9</sup>

Growth specification ( $T = \#$  year lags)

$$\overbrace{\ln(Y_{it}/Y_{it-T})}^{\text{Output growth}} = \overbrace{\alpha \ln(K_{it}/K_{it-T}) + \beta \ln(L_{it}/L_{it-T})}^{\text{Growth in use of inputs}} + \underbrace{\tau T + \beta/\rho \ln(\Omega_{it}/\Omega_{it-T})}_{\text{TFP growth}} + \omega_{it} - \omega_{it-T} \quad [3]$$

with

$$\Omega_{it} \equiv S_{it}^r + \lambda_{1r} [S_{it}^1]^\rho + \dots + \lambda_{nr} [S_{it}^n]^\rho$$

$S_{it}^j \equiv L_{it}^j / L_{it}$  the employment shares  $j=1 \dots n$

$\lambda_{jr} \equiv \mu_j / \mu_r$ ;  $j=1 \dots n, j \neq r$ ;  $r = \text{ref. cat.}$

and rel. (marginal) labour productivities (dropping  $t$ )

$$\partial Y / \partial L^j / \partial Y / \partial L^r = \mu_j / \mu_r (L^j / L^r)^{\rho-1} = \lambda_{jr} (S^j / S^r)^{\rho-1}$$

# Data and descriptive statistics

## Source 1: Bel-first ( $Y_{it}, K_{it}\dots$ )

- Panel of about 9,000 firms (>20 workers) located in Belgium, from all sectors of the for-profit economy (except agri & mining), from 1998 to 2006
- Info on sector, location, size, capital ( $K_{it}$ ), labour & value added ( $Y_{it}$ ), edu. attainment, ownership nationality, multinational status

## Source 2: Carrefour database (i.e. social security registers) ( $S_{it}^j$ )

- Individual-level information on age, gender, blue-/white collar status of all workers from Bel-first sample.

- Aggregation of Carrefour data at firm level + merge with Bel-first
- Resulting firm-level panel contains labour shares  $S_{it}^j ; j \equiv Age \times Gender \times Blue/White \text{ collar status}$

## NB about educational attainment

- Educational attainment (primary, secondary, tertiary degree) only available at firm-level in Bel-first; while age, gender & blue/white-collar status exists at individual level in Carrefour
- We proxy education using the white/blue collar status and interact it with the two other dimensions
- But we provide evidence that this approximation is reasonable

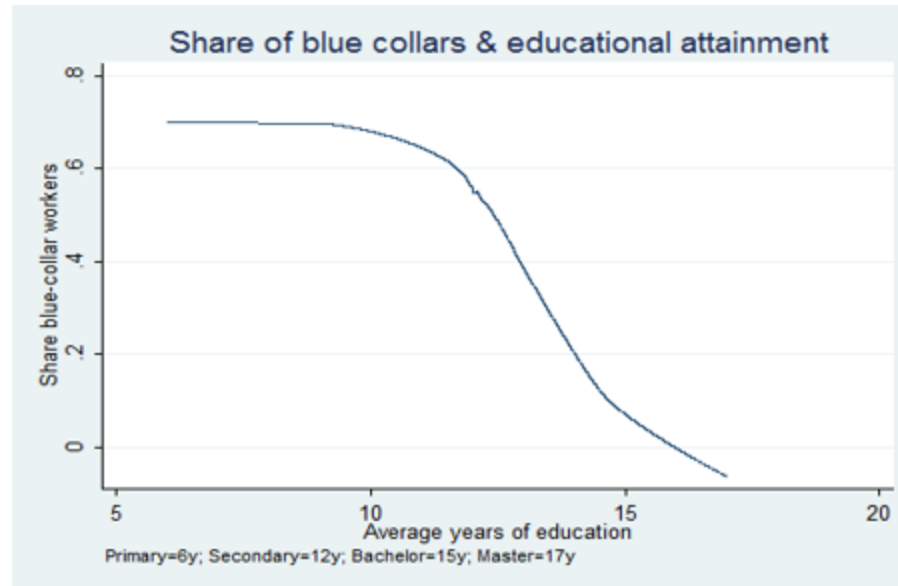
*Table 4: Ageing-Feminisation and rising educational attainment*

	Age-mean	Share female	Share blue collars	Aver. years education*
1998	36.567	0.249	0.563	11.490
1999	36.609	0.256	0.551	11.562
2000	36.695	0.262	0.541	11.631
2001	36.764	0.271	0.529	11.709
2002	37.336	0.280	0.488	11.769
2003	37.873	0.281	0.482	11.818
2004	38.109	0.284	0.481	11.766
2005	38.363	0.289	0.475	11.816
2006	38.689	0.294	0.465	11.803
<i>N</i>	75,437			

Source: Bel-first; Carrefour. Weight: number of fte (full-time equivalent) workers in the firm.

\*Primary degree=6; Secondary degree=12, Bachelor=15 and Master=17 years

Figure 3- Share of blue-collar workers & average educational attainment. Year 2006.



Source: Bel-first; Carrefour. Weight: number of fte workers in the firm. Based on lowess estimation i.e. locally weighted regression of  $y$  on  $x$ .

# Econometric results



Table 5 - Econometric analysis of the role of age(ing) on & TPF level and growth- 7 age groups:<30,30-35,35-40[ref],40-45,45-50,50-55,55-65

	[1] Level	[2] Growth(FE)	[3] Growth(FE)+ controls	[4] Growth(FE)+ controls incl. cohorts	[5] Growth(FE)+ controls incl. cohorts/2steps LP
<i>Cst</i>	4.110*** (0.0266)	0.0377*** (0.00115)	0.0280*** (0.00400)	0.0488** (0.0180)	0.0684*** (0.0199)
$\alpha$	0.112*** (0.00119)	0.0317*** (0.00271)	0.0423*** (0.00313)	0.0423** (0.0148)	
$\beta$	0.908*** (0.00250)	0.638*** (0.00487)	0.574*** (0.00583)	0.571*** (0.0244)	0.273*** (0.0677)
$\rho$	1.054*** (0.0163)	0.792*** (0.0128)	0.794*** (0.0169)	0.790*** (0.0383)	0.540*** (0.0710)
$\eta_{<30}$ (a)	-0.599***	0.079**	0.217***	0.187**	0.263*
$\eta_{30-34}$	0.209***	-0.026	0.010	-0.013	-0.087
$\eta_{40-44}$	-0.212***	-0.144***	-0.066	-0.039	-0.140
$\eta_{45-49}$	-0.537***	-0.237***	-0.116**	-0.057	-0.210*
$\eta_{50-54}$	-0.566***	-0.360***	-0.318***	-0.236***	-0.436***
$\eta_{54-64}$	-0.143***	-0.376**	-0.396***	-0.277**	-0.580***
Controls	Year*Sector	Firm fixed effects	Firm fixed effects+ Share of women, blue-collar wks	Firm fixed effects+ Share of women, blue-collar wks + cohort	Firm fixed effects+ Share of women, blue-collar wks + cohort
Nobs	75,437	65,750	48,777	48,777	48,076
$\sigma \equiv 1/(1-\rho)$	-18.643	4.810	4.865	4.751	2.172
<i>Implied relative marginal productivities (1=35-39 ref)</i>					
<i>RMP</i> <sub>&lt;30</sub>	0.403	1.014	1.151	1.121	1.156
<i>RMP</i> <sub>30-34</sub>	1.192	0.982	1.025	1.002	0.965
<i>RMP</i> <sub>35-39</sub>	1(ref)	1(ref)	1(ref)	1(ref)	1(ref)
<i>RMP</i> <sub>40-44</sub>	0.771	0.880	0.954	0.983	0.923
<i>RMP</i> <sub>45-50</sub>	0.449	0.812	0.935	0.999	0.920
<i>RMP</i> <sub>50-55</sub>	0.417	0.701	0.744	0.835	0.704
<i>RMP</i> <sub>55-65</sub>	0.814	0.699	0.674	0.809	0.554

Standard errors in parentheses All models are estimated using non-linear least squares, with standard errors robust to firm-level clustering. Source: Bel-first; Carrefour \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

(a):  $\eta \equiv \lambda - 1$

Table 6 - Age-gender & TFP growth- 7 age groups:<30,30-34,35-39,40-44,45-49,50-54,55-64

		[3] Growth(FE) + controls		
	<i>Cst</i>	0.0713*** (0.0177)		
	$\alpha$	0.0266*** (0.00517)		
	$\beta$	0.515*** (0.0272)		
	$\rho$	0.774*** (0.0335)		
		Women	Men	Prob $\eta_j$ W=M
	$\eta_{<30}$ (a)	-0.112	0.181**	0.001***
	$\eta_{30-34}$	-0.170	-0.048	0.242
	$\eta_{35-39}$	-0.240	0 (ref)	0.048**
	$\eta_{40-44}$	-0.246**	-0.048	0.064*
	$\eta_{45-49}$	-0.218*	-0.149*	0.591
	$\eta_{50-54}$	-0.239**	-0.248***	0.950
	$\eta_{55-64}$	-0.432***	-0.202*	0.290
	Controls	Firm fixed effects+ Share of part-time workers, blue-collar wks		
	Nobs	40,969		
	$\sigma \equiv 1/(1-\rho)$	4.432		
		<i>Implied relative marginal productivities (1=35-39 ref)</i>		
	<i>RMP</i> <sub>&lt;30</sub>	0.925	1.117	
	<i>RMP</i> <sub>30-34</sub>	0.905	0.975	
	<i>RMP</i> <sub>35-39</sub>	0.835	1 (ref)	
	<i>RMP</i> <sub>40-44</sub>	0.833	0.970	
	<i>RMP</i> <sub>45-49</sub>	0.875	0.896	
	<i>RMP</i> <sub>50-54</sub>	0.868	0.814	
	<i>RMP</i> <sub>55-64</sub>	0.654	0.883	

Standard errors in parentheses. All models are estimated using non-linear least squares, with standard errors robust to firm-level clustering. Source: Bel-first 1998-2006; Carrefour. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

(a):  $\eta \equiv \lambda - 1$

Table 10 – Age, blue/white collar status & TPF growth- 7 age groups:<30,30-35,35-40,40-45,45-50,50-55,55-65

AGE x Blue vs white collars

	Growth(FE) + cohorts		
$Cst$	0.0826*** (0.0189)		
$\alpha$	0.0250*** (0.00590)		
$\beta$	0.560*** (0.0248)		
$\rho$	0.856*** (0.0319)		
	Blue collars	White collars	Prob $\eta_{blue=white}$
$\eta_{<30(a)}$	0.019	-0.101	0.238
$\eta_{30-34}$	-0.119	-0.110	0.928
$\eta_{35-40}$	-0.081	0 (ref)	0.395
$\eta_{45-49}$	-0.213**	-0.236**	0.814
$\eta_{50-54}$	-0.331***	-0.282***	0.681
$\eta_{54-64}$	-0.391***	-0.479***	0.524
$\eta_{54-64}$	-0.275*	-0.604**	0.065*
Controls	Firm fixed effects+ Share of part-time workers, blue-collar wks		
Nobs	47830		
$\sigma \equiv 1/(1-\rho)$	6.947		
<i>Implied relative marginal productivities (1=35-39 ref)</i>			
$RMP_{<30}$	0.903	0.873	
$RMP_{30-34}$	0.824	0.888	
$RMP_{40-44}$	0.850	1 (ref)	
$RMP_{45-50}$	0.732	0.781	
$RMP_{50-55}$	0.633	0.752	
$RMP_{55-65}$	0.591	0.551	
$RMP_{<30}$	0.729	0.418	

Standard errors in parentheses. All models are estimated using non-linear least squares, with standard errors robust to firm-level clustering. Source: Bel-first 1998-2006; Carrefour. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

(a):  $\eta \equiv \lambda - 1$

Table 8 - Age (<30,30-49,50-64), gender, blue/white collar status & TFP growth

AGE x  
GENDER X  
Blue/White  
collars

[3] Growth(FE) + controls								
Cst	0.0814*** (0.0168)							
$\alpha$	0.0255*** (0.00575)							
$\beta$	0.564*** (0.0240)							
$\rho$	0.861*** (0.0317)							
	Blue collars			White collars			Prob $\eta_j$ Blue=White	
	Women	Men	Prob $\eta_j$ W=M	Women	Men	Prob $\eta_j$ W=M	M	W
$\eta_{<30}$ (a)	-0.076	0.045	0.241	-0.201	-0.019	0.110	0.502	0.349
$\eta_{30-49}$	-0.276*	0.009	0.067*	-0.187	0 (ref)	0.184	0.927	0.610
$\eta_{50-64}$	-0.354**	-0.207*	0.455	-0.489***	-0.335**	0.458	0.431	0.553
Controls	Firm fixed effects+ Share of part-time workers							
Nobs	50,398							
$\sigma \equiv 1/(1-\rho)$	7.180							
Implied relative marginal productivities (1=30-49 white collar man ref)								
	Blue collars			White collars				
	Women	Men		Women	Men			
$RMP_{<30}$	0.904	0.933		0.814	1.003			
$RMP_{30-49}$	0.718	0.891		0.847	1 (ref)			
$RMP_{50-64}$	0.681	0.775		0.559	0.715			

Standard errors in parentheses

All models are estimated using non-linear least squares, with standard errors robust to firm-level clustering.

Source: Bel-first; Carrefour 1998-2006

\*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

(a):  $\eta \equiv \lambda - 1$

## To sum up

- Strong (negative) effect of age on TFP growth
- No statistically significant additional impact of:
  - gender
  - blue-collar status → education does not counterbalance the negative effect of ageing

## Additional results:

- Industry (service VS manufacturing) : age-related decline of productivity is slightly more pronounced in manufacturing
- Region (VL, W, Bx1) : no differential effect
- Foreign Ownership : no differential effect
- Multinational Status: no differential effect

# Estimating the overall impact of ageing on TFP growth ...

1990-2040

## i) Strategy & data

The key idea is to use

- estimated  $\hat{\lambda}$ 's,  $\hat{\beta}$  and  $\hat{\rho}$  (stemming from the estimation of model [3] using 1998-2006 firm-level data)
- alongside observed + projected values of the labour shares by age  $S_t^j$ ,

to compute

$$\text{Annual TFP growth loss} = \hat{\beta} / \hat{\rho} \ln (\Omega_t / \Omega_{t-1}) \quad [4]$$

$$\text{with } \Omega_t \equiv [S_t^r]^{\hat{\rho}} + \hat{\lambda}_{1r} [S_t^1]^{\hat{\rho}} \dots + \hat{\lambda}_{nr} [S_t^n]^{\hat{\rho}}$$

NB : we drop firm index  $i$  as we no longer work with firm-level Bel-first data

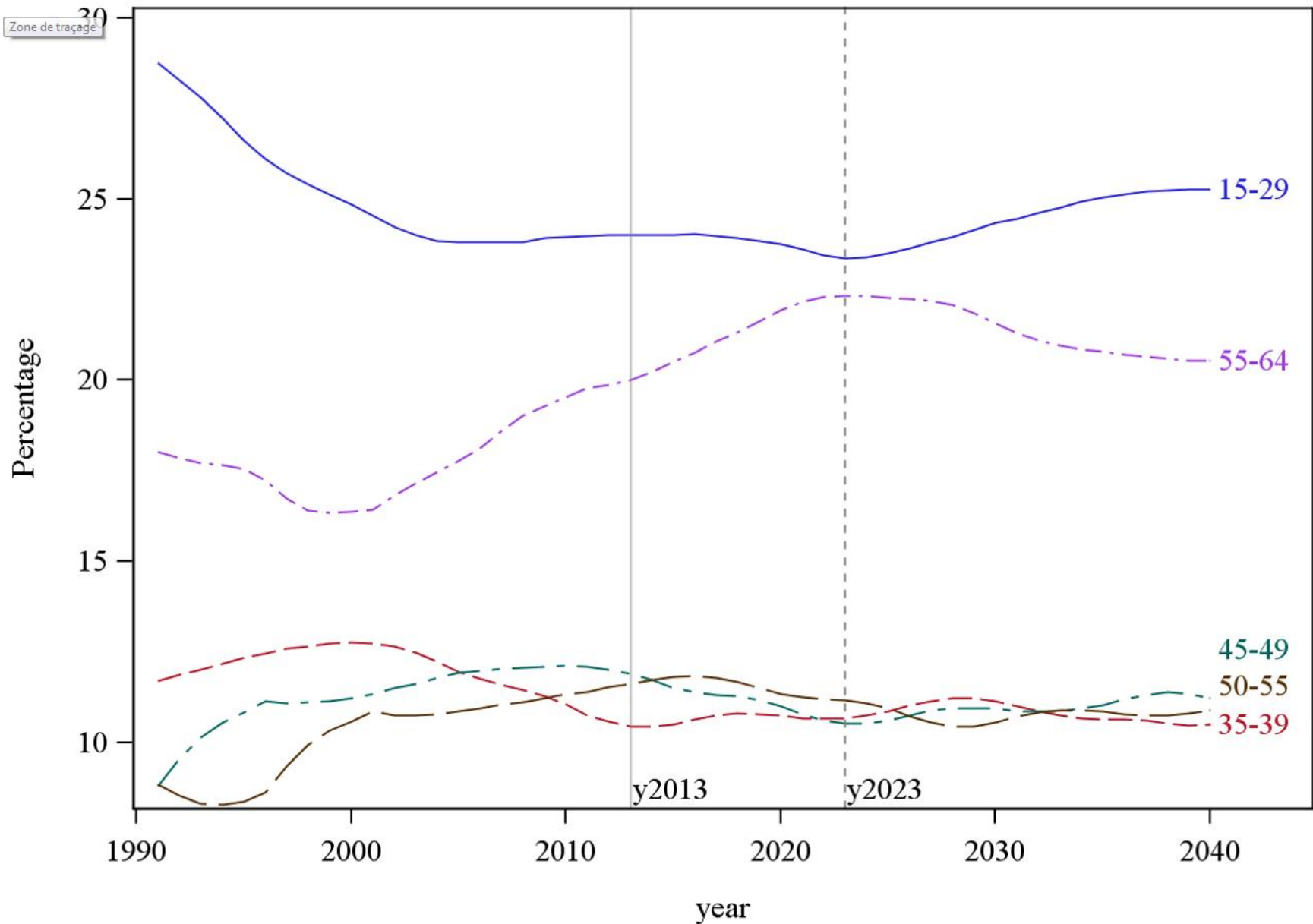


Table 5 - Econometric analysis of the role of age(ing) on & TPF level  
 34, 35-39[ref], 40-44, 45-49, 50-54, 5!

	[1] Level	[2] Growth(FE)	[3] Growth(FE)+ controls
<i>Cst</i>	4.110*** (0.0266)	0.0377*** (0.00115)	0.0280*** (0.00400)
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$\eta_{45-49}$	-0.537***	-0.237***	-0.116**
$\eta_{50-54}$	-0.566***	-0.360***	-0.318***
$\eta_{55-64}$	-0.143***	-0.376**	-0.396***

(a):  $\eta \equiv \lambda - 1$

# ii) Evolution population shares by age



Source: INS 2014, population perspectives 2013-2060

### iii) From population shares to employment shares

This said, demographics ( $P_t^j$ ) is only one part of the full story. What matters are employment shares ( $S_t^j$ ), driven by (relative) employment rates ( $ER_t^j$ )

$$S_t^j = (ER_t^j / ER_t) P_t^j \quad [5]$$

- For the period 1991-2013, employment shares are known
- Beyond 2013, assumptions about employment rates  $ER_t^j$  are needed

S 1= we freeze employment rates ( $ER_t^j$ ) to their 2013 levels

S 2= EU target of a 75% overall employment rate, in 2020

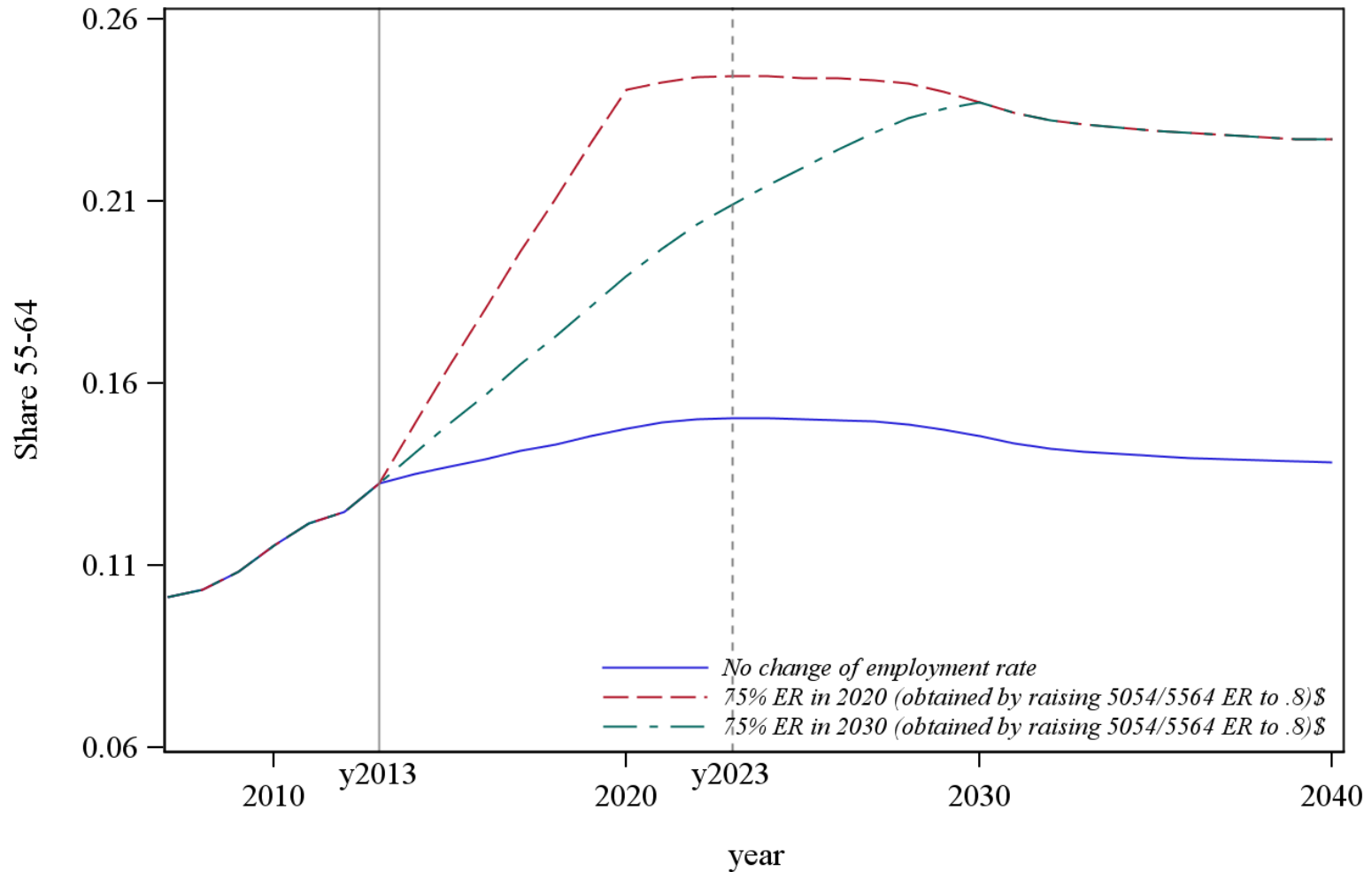
S 3= EU target of a 75% -----in 2030

Table 12: Hypothesis about evolution of employment rates by age (2014-2040)

Scenario 1	<30	30-34	35-39	40-44	45-49	50-54	55-64
2013	.414	.803	.815	.817	.805	.751	.417
2020	.414	.803	.815	.817	.805	.751	.420
2030	.414	.803	.815	.817	.805	.751	.420
2040	.414	.803	.815	.817	.805	.751	.420
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Scenario 2	<30	30-34	35-39	40-44	45-49	50-54	55-64
2013	.414	.803	.815	.817	.805	.751	.417
2020	.414	.850	.850	.850	.850	.800	.800
2030	.414	.850	.850	.850	.850	.800	.800
2040	.414	.850	.850	.850	.850	.800	.800
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Scenario 3	<30	30-34	35-39	40-44	45-49	50-54	55-64
2013	.414	.803	.815	.817	.805	.751	.417
2020	.414	.822	.829	.831	.824	.771	.576
2030	.414	.850	.850	.850	.850	.800	.800
2040	.414	.850	.850	.850	.850	.800	.800

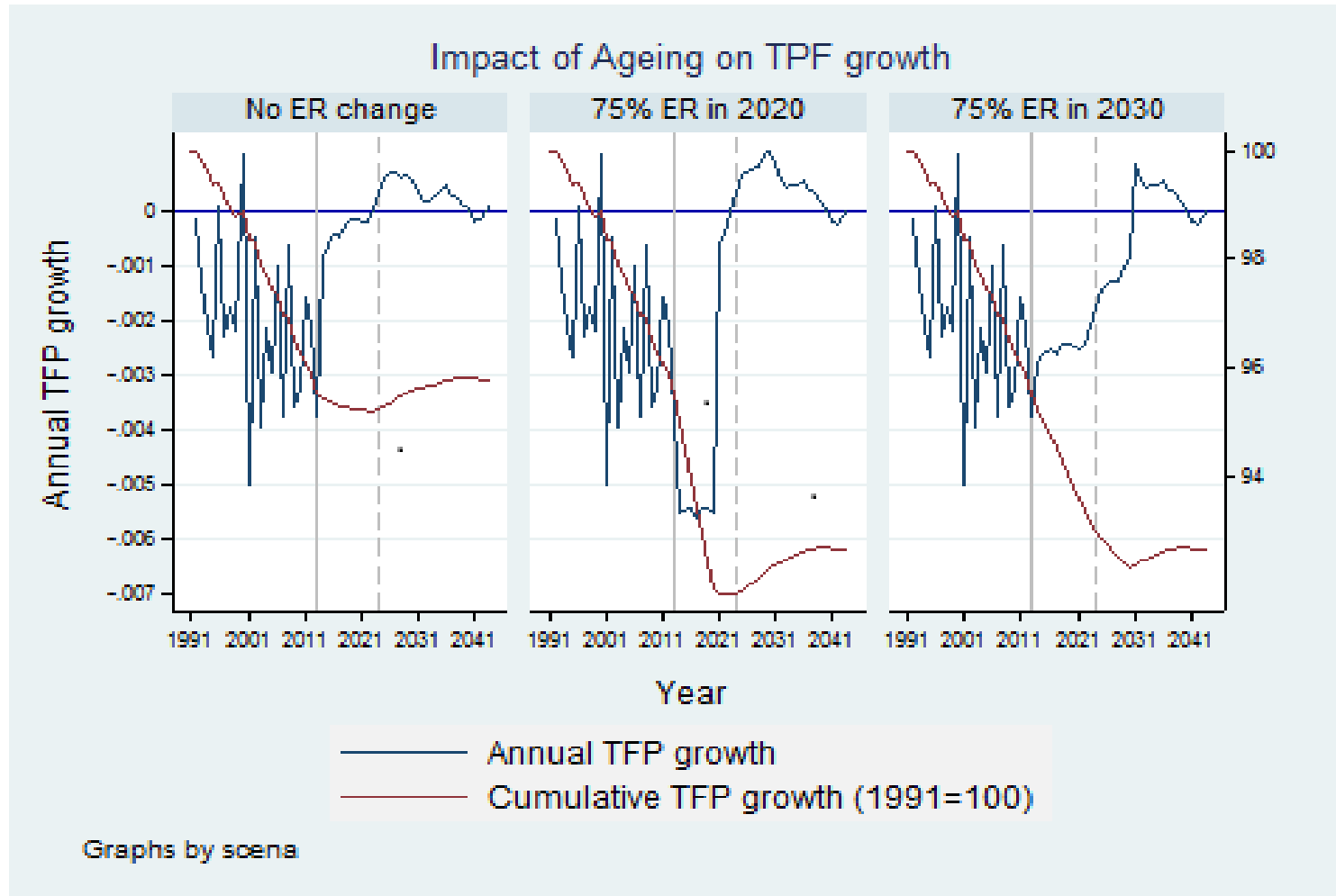
Source: OECD-LFS, our calculus

### Predicted evolution of employment share (55-64)

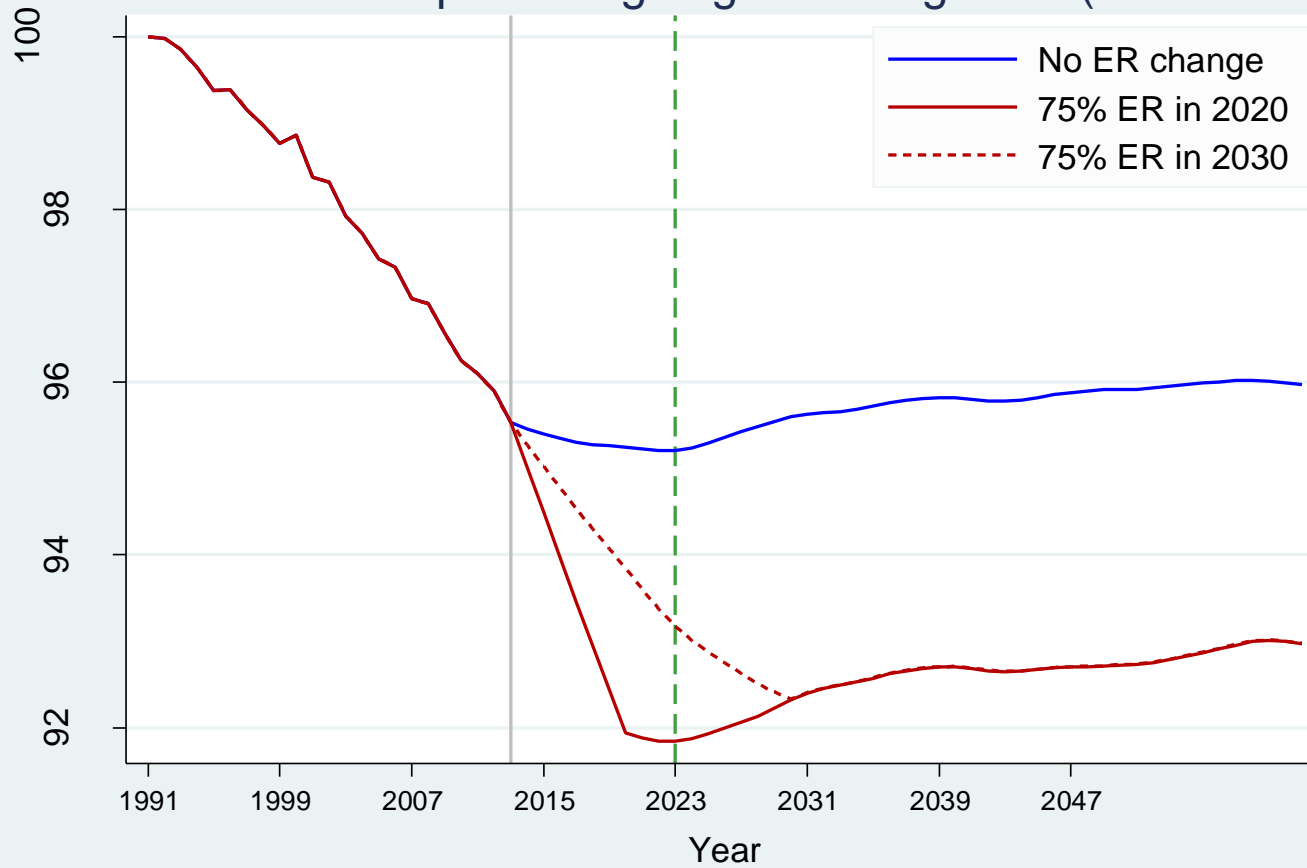


\$.EU2020 target of 75% of the 20-64 year-olds to be employed  
 Source: INS 2014, Population Perspectives 2013-2060, OECD, Eurostat. Our calculs

Figure 7a,b – Impact of ageing on annual and cumulative TFP growth: 1991-2040, three scenari



### Cumulative impact of ageing on TFP growth (1991=100)



# Conclusions

This paper examines the role of socio-demographic changes in the composition of the workforce on TFP growth

- Ageing workforce
- Feminization of ageing
- Increased educational attainment of older workers.

Using data on Belgian firms (1998-2006), we find evidence that the TFP growth slowdown could have been driven by ageing

- But no gender bias
- And no sign that rising educational attainment could counterbalance ageing
- The impact of ageing uniform across industries, regions & degrees of international exposure



Combining firm-level results & country-level demographic/employment data, we estimate that

- over the 1991-2013 period, ageing may have dented cumulative TPF growth by -4.5 percentage points.
- that loss could rise to -7 percentage points by 2030

The latter is not so much dictated by Belgium's demography (peak of ageing workforce = mid-2020s)...

Rather by the EU target of 75% overall employment rate. Reaching that target by 2020/2030 will require a sharp rise of the 55-64 employment rate

# Thank you!

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