

# You're the one that I want! Understanding the over-representation of women in the public sector

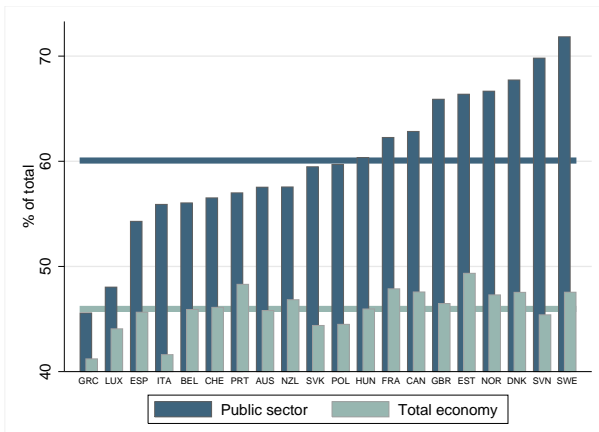
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*EEA Meetings-Rotterdam*

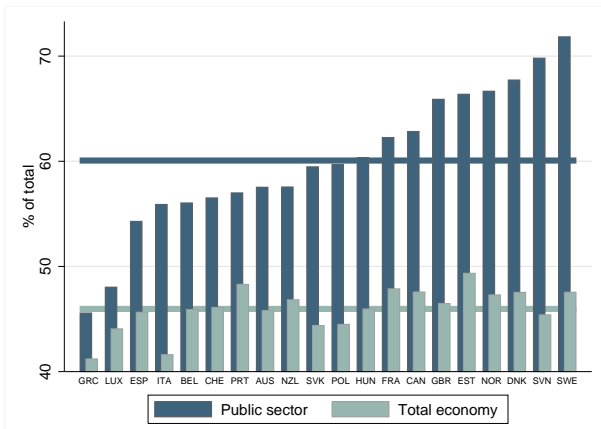
August 27th, 2024

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→ Robust to measurement, over time, age, education, industries, occupations..

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- ③ better work-life balance;
- ④ higher job-security.



# Lower gender wage gaps in the public sector

$$\log(y_i^a) = \beta_0 + \beta_1 f + \beta_2 X_i + \beta_3 m \times pub + \beta_4 f \times pub + \beta_5 C_i + d_r + d_y + \epsilon_i$$

Note: CPS for 1996-2018, Structure of Earnings Survey from 2002, 2006, 2010, and 2014.



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	US		Spain	
	(1)	(2)	(1)	(2)
<b>Public-sector wage premium</b>				
Men	-0.0198*** (-6.26)	-0.0289*** (-9.48)	0.0015 (0.62)	0.0082*** (3.68)
Women	0.0547*** ( 19.67)	0.0530*** ( 18.63 )	0.0688*** (28.65)	0.0732*** (32.43)
<b>Gender wage gap</b>				
Private	-0.2888*** (-158.68)	-0.2861*** (-156.20)	-0.2145*** (-163.18)	-0.2472*** (-196.78)
<b>Controls</b>				
Age	X	X	X	X
Education	X	X	X	X
Reg., year	X	X	X	X
Tenure <sup>2</sup>			X	X
Occupat.	X	X	X	X
Part time dummy	X		X	
Only full time wkr		X		X
Obs.	1,037,822	796,920	876,274	747,228
R-squared	0.5090	0.3975	0.5989	0.5383

Note: CPS for 1996-2018, Structure of Earnings Survey from 2002, 2006, 2010, and 2014.

# Fewer annual working hours in the public sector

$$\log(hours_i) = \alpha_0 + \alpha_1 f + \alpha_2 X_i + \alpha_3 pub + \alpha_4 C_i + d_r + d_y + \epsilon_i$$

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	US		Spain	
	(1)	(2)	(1)	(2)
<b>Panel B: hours regressions</b>				
<b>Public sector hours premium</b>				
Public	-0.0109*** (-10.15)	-0.0286*** (-37.00)	-0.0419*** (-38.11)	-0.0365*** (-74.59)
<b>Controls</b>				
Age	X	X	X	X
Education	X	X	X	X
Reg., year	X	X	X	X
Tenure <sup>2</sup>			X	X
Occupat.	X	X	X	X
Part time dummy	X		X	
Only full time wkr		X		X
Obs.	1,008,225	777,538	876,274	747,228
R-squared	0.3799	0.0685	0.5478	0.3031

Note: CPS for 1996-2018, Structure of Earnings Survey from 2002, 2006, 2010, and 2014

Hazard rates by sector and gender

	US			Spain		
	All	Men	Women	All	Men	Women
$P \rightarrow U$	<b>0.015</b>	0.016	0.012	<b>0.043</b>	0.042	0.044
$P \rightarrow I$	0.023	0.019	0.027	0.033	0.025	0.045
$G \rightarrow U$	<b>0.007</b>	0.006	0.008	<b>0.022</b>	0.020	0.023
$G \rightarrow I$	0.018	0.015	0.020	0.024	0.019	0.027

▶ Conditional probabilities

- $P$  = Private sector
- $G$  = Public sector

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# Ingredients for a model

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- ② Better work-life balance:
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- ③ Job security:
  - Flows from public (private) employment to unemployment
- ④ Preferences:
  - Residual needed to explain female over-representation.

# Model: General setting

- Search and matching model with .
  - Men and women  $j = [m, f]$ .
  - Private and public sector  $i = [g, p]$ ,
- Continuous time.
- Risk neutral agents with discount rate  $r > 0$ .

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  - Men and women  $j = [m, f]$ .
  - Private and public sector  $i = [g, p]$ ,
- Continuous time.
- Risk neutral agents with discount rate  $r > 0$ .
- At each instant,  $\tau$  individuals die and a new generation is born.
- Draw a preference for public sector  $\epsilon_j$ , distributed  $\Xi^j(\cdot)$  on  $[-\infty, \infty]$ .
  - Join the public-sector labor market if  $\epsilon$  is high enough.
  - Join the private sector otherwise.
- In each market, people can be employed (e) or non-employed (ne).

- At rate  $\lambda$ , they draw  $x$ , value of home production
  - Men draw from  $F_m(\cdot)$ ,
  - Women draw from  $F_f(\cdot)$  [different by gender].
- Flow utilities:

$$v_{i,j}^E = (1 - \xi_i)x + w_{i,j},$$
$$v_{i,j}^{NE} = x,$$

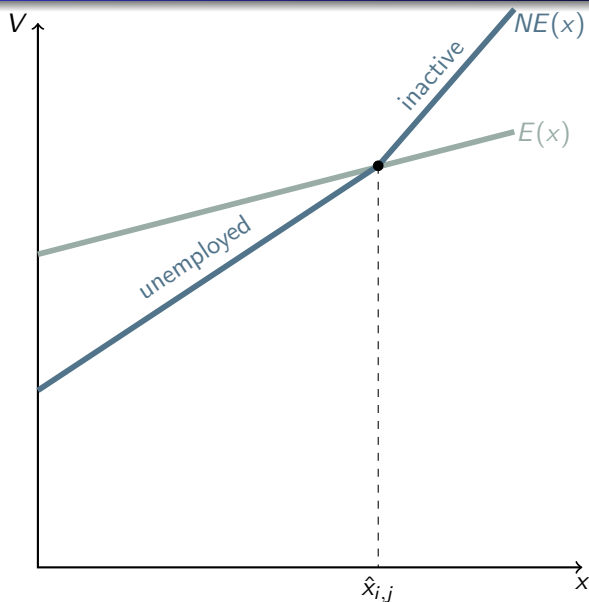
- Time costs of work  $\xi_i$  [different across sectors].

$$(r + \tau + \lambda)E_{i,j} = v_{i,j}^E + \delta_i [NE_{i,j} - E_{i,j}] + \lambda \int_0^\infty \max(E_{i,j}(x'), NE_{i,j}(x')) dF_j(x')$$

$$(r + \tau + \lambda)NE_{i,j} = v_{i,j}^{NE} + m(\theta_i) [\max(E_{i,j}, NE_{i,j}) - NE_{i,j}] + \lambda \int_0^\infty NE_{i,j}(x') dF_j(x')$$

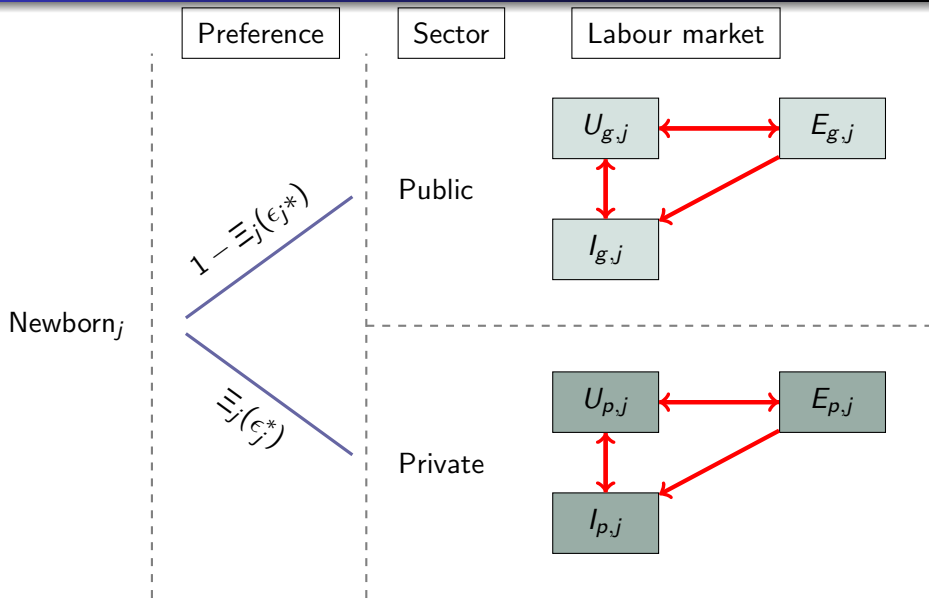
- Exogenous separations at rate  $\delta_i$  [different across sectors].

# Decision Threshold



▶ Threshold

# Decision of newborn - Segmented markets



► Flows between States

# Closing the model

- Private sector:
  - Male wage:  $w_m = \beta y$
  - Value of a job same for male and female workers:  $J_m = J_f$
  - Endogenous gender wage gap: Women are more likely to quit.
  - To target observed gap: Assume a “wedge” on women’s wages
  - Free-entry condition pins down  $\theta_p$  and hence  $m(\theta_p) = \zeta \theta_p^\eta$ .



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- Government:
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  - Hires workers to compensate for the ones who retire, or separate into unemployment or inactivity.
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  - Hires workers to compensate for the ones who retire, or separate into unemployment or inactivity.
  - **Does not react to market tightness.**
- Initial choice of sector:
  - Comparing expected values of inactivity and unemployment in the private and the public sector, including preferences  $\epsilon_j$ :
  - $\max \{ (1 - F_j(\hat{x}_{p,j}))I_{p,j} + F_j(\hat{x}_{p,j})U_{p,j}; (1 - F_j(\hat{x}_{g,j}))I_{g,j} + F_j(\hat{x}_{g,j})U_{g,j} + \epsilon_j \}$

# Summarizing the model

Differences between men and women:

- Value of home production:  $F_j(\cdot)$ ,
- Preference for sector:  $\Xi^j(\cdot)$ ,
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Differences between sectors:

- Time cost:  $\xi_i$ ,
- Exogenous separation rate:  $\delta_i$ ,
- Wages:  $w_{i,m}$ ,  $w_{i,f}$ ,
- Job creation:  $P$  responds to tightness;  $G$  does not.

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Interaction  $P$  &  $G$  : Public-sector employment and wages affect:

- expected values for searching in the public sector.
- decision to join each sector.
- tightness and job creation in the private sector.

# Mechanisms behind over-representation

- Lower gender wage gap public sector, higher premium  $\pi_f$ 
  - Value for women's employment in public sector increases.
  - On the margin, men prefer a “less crowded” private sector.
  - Magnitude of effect depends on preference distribution.

# Mechanisms behind over-representation

- Better conciliation, less working time (lower  $\xi_g$ )
  - Value for employment in public sector increases for both genders.
  - If women have on average higher opportunity costs of working, effect will be stronger for them.
  - Again, in that case men would prefer a “less crowded” private sector.

# Mechanisms behind over-representation

- Higher job security, lower job separation rate  $\delta_g$ 
  - Value for employment and non-employment in public sector increases.
  - If women have higher opportunity costs of working and lower wages, might benefit less from safer jobs.
  - In that case, women might prefer “less crowded” private sector.



# Mechanisms behind over-representation

- Different preferences
  - Different means affect gender composition across sectors on top of job characteristics; variance determines magnitude of crowding out effect.

# Calibration strategy

- France, UK, Spain, US separately.
- Fix 11 parameters.
- Calibrate 9 parameters.
- 9 targets



# Drivers for women's over-representation

Benchmark	No sector differences no preference differences	No wage difference	No hours diff.	No job security diff	No sector diff.	
	$\pi_j = 1$ $\xi_g = \xi_p$ $\delta_g = \delta_p$ $\bar{e}_f = \bar{e}_m$	$\pi_j = 1$	$\xi_g = \xi_p$	$\delta_g = \delta_p$	$\pi_j = 1$ $\xi_g = \xi_p$ $\delta_g = \delta_p$	
Public-sector employment shares ratio						
US	1.36	0.99	1.15( <b>57.1%</b> )	1.35(1.9%)	1.36(-1.2%)	1.16(56.4%)
UK	1.86	1.00	1.86(-0.9%)	1.83(2.4%)	1.87(-1.3%)	1.87(-2.3%)
FR	1.59	1.01	1.53(10.5%)	1.55(6.9%)	1.60(-2.2%)	1.48( <b>18.3%</b> )
ES	1.56	0.93	1.33(40.1%)	1.52(6.7%)	1.58(-4.2%)	1.33( <b>41.2%</b> )
US						
College	1.41	0.89	0.90( <b>124.6%</b> )	1.45(-9.6%)	1.41(-0.8%)	1.02(96.2%)
Non-college	1.19	1.00	1.19(2.3%)	1.19(1.6%)	1.19(-1.6%)	1.19(2.1%)

Alternative decomposition: 

# Work life balance and job-security premia

Percentage of wage private sector workers would sacrifice for public sector hours

Country	$[\xi_p = \xi_g]$	
	Women	Men
US	0.84	0.81
UK	1.46	0.93
France	3.52	2.82
Spain	2.23	1.78

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Percentage of wage private sector workers would sacrifice for public sector separation rates

Country	$[\delta_p = \delta_g]$	
	Women	Men
US	0.98	1.02
UK	0.92	1.17
France	1.43	1.66
Spain	2.88	3.39

# Effects of public-sector policies for Spain

<b>Policy</b>	<b>Spain</b>
<i>Increase of wages by 1 percent</i>	
$\Delta$ unemployment rate male	0.07 pp.
$\Delta$ unemployment rate female	0.15 pp.
$\Delta$ inactivity rate male	-0.06 pp.
$\Delta$ inactivity rate female	-0.09 pp.
$\Delta$ aggregate wage gap	-0.05 pp.
<i>Increase of employment by 1 percent</i>	
$\Delta$ unemployment rate male	-0.03 pp.
$\Delta$ unemployment rate female	-0.06 pp.
$\Delta$ inactivity rate male	0.01 pp.
$\Delta$ inactivity rate female	0.01 pp.
$\Delta$ aggregate wage gap	-0.01 pp.

▶ Other countries

- 1 Selection of women into the public sector is driven by:
  - an intrinsic preference for public-sector occupations (mainly in the UK).
  - lower gender wage gaps (57% in US, 40% Spain, 11% in France);
  - better work-life balance (7% in France and Spain);
  - no role for job security;
  - stark differences across educational groups.

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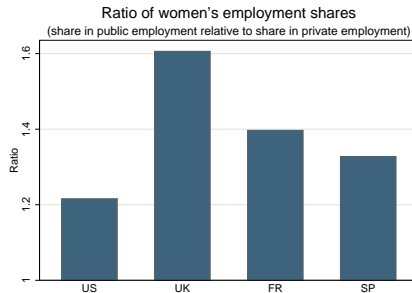
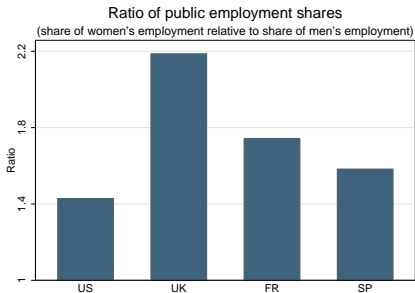


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  - the public-sector job-security premia is 1 to 4 percent - higher for men;
  - the public-sector work-life balance premia is 1 to 4 percent - higher for women.
- 3 Policies regarding public sector wages and employment affect women's unemployment 2-3 times as much as men's.

# Gender bias in public employment

.. is robust to measurement



# Public-sector employment, by gender

	Public sector	Private sector	Total
Women	$e_{g,f}$	$e_{p,f}$	$e_f$
Men	$e_{g,m}$	$e_{p,m}$	$e_m$
Total	$e_g$	$e_p$	$e$

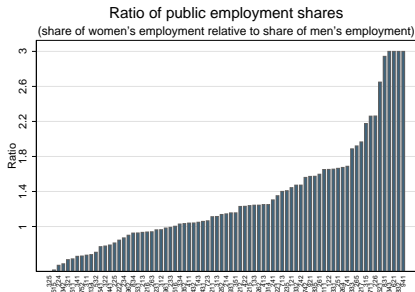
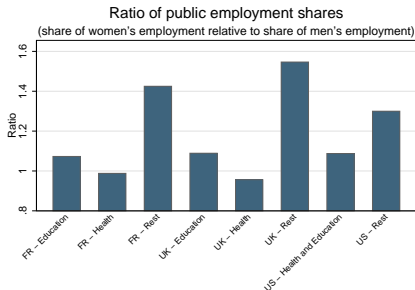
Note: Government ( $g$ ), private ( $p$ ), women ( $f$ ), men ( $m$ ).

Two statistics:

- Ratio of public employment share:  $rg = \frac{\frac{e_{g,f}}{e_f}}{\frac{e_{g,m}}{e_m}}$ ,
- Ratio of women employment shares :  $rf = \frac{\frac{e_{g,f}}{e_g}}{\frac{e_{p,f}}{e_p}}$ ,

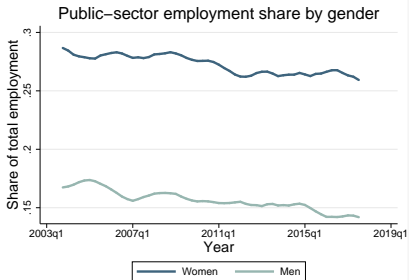
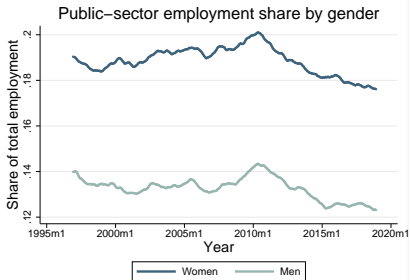
# Gender bias in public employment

...across industries and occupations..

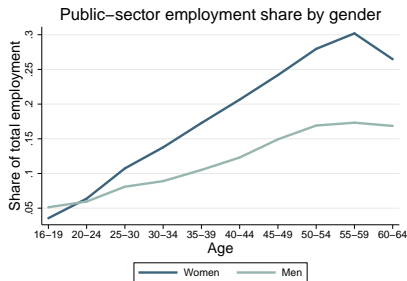
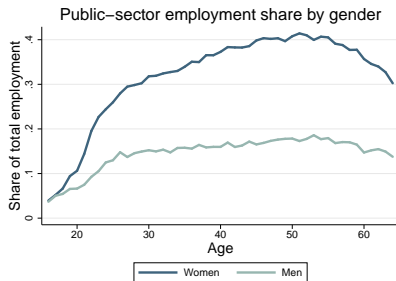
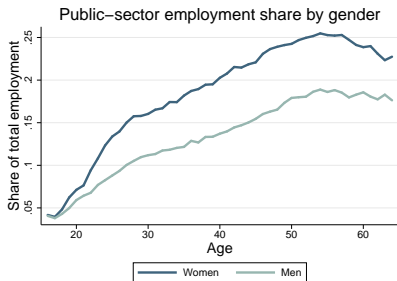


Note: French, UK Labour Force Surveys, CPS. For occupations: CPS data, average between 1996 and 2017. 3-digit occupations that have an overall share of public-sector employment between 0.05 and 0.95.

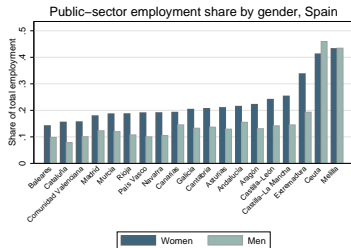
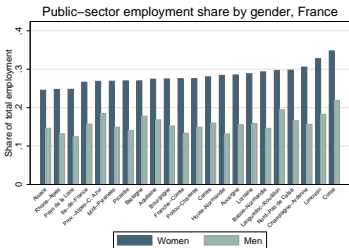
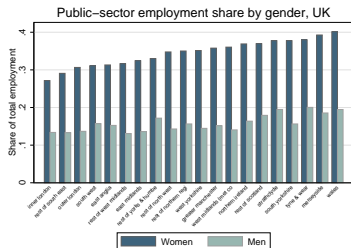
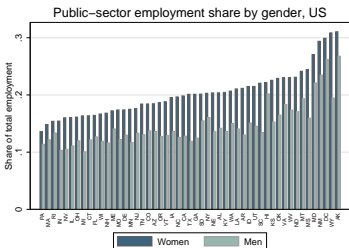
# Public-sector employment by gender over time



# Public-sector employment by gender, different age groups



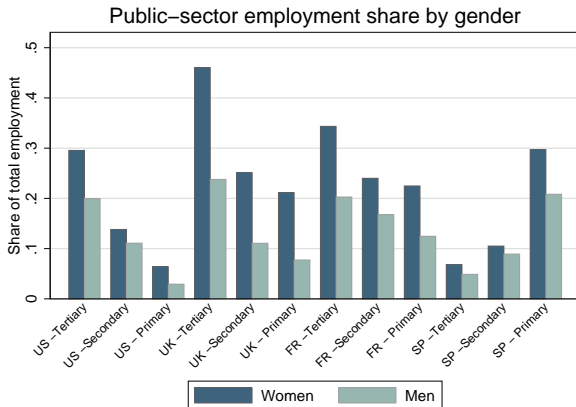
# Public-sector employment by gender and regions



Note: French, Spanish, and UK Labour Force Surveys and CPS (2003-2018).

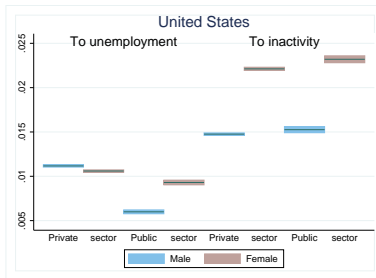
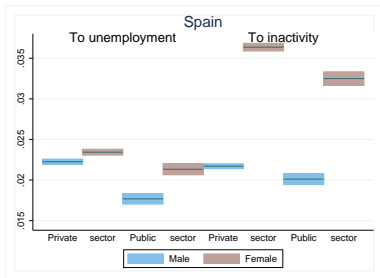


# Public-sector employment by gender, educational groups



*Note: For the United States the data is taken from the CPS (2003-2018), for the United Kingdom from the British Household Panel Survey (2003-2018), for France from the French Labour Force Survey (2003-2017) and for Spain from the Spanish Labour Force Survey (2003-2018).*

# Conditional job-separations: US and Spain



**Multinomial logit model:** estimate the probabilities of transitioning out of employment conditional on observable characteristics.

$$\lambda_i^U = \frac{\exp(x_i\beta_U)}{1 + \exp(x_i\beta_U) + \exp(x_i\beta_I)}, \quad (1)$$

$$\lambda_i^I = \frac{\exp(x_i\beta_I)}{1 + \exp(x_i\beta_U) + \exp(x_i\beta_I)}, \quad (2)$$

$x_i$  includes age, age squared, and dummies for education, region, year, public sector, occupation, age between 60-64, gender and gender dummies interacted with being employed in the public sector. [Return](#)

# Steady-state equilibrium

A set of thresholds  $\{\bar{e}_f, \bar{e}_m, \hat{x}_{g,m}, \hat{x}_{p,m}, \hat{x}_{g,f}, \hat{x}_{p,f}\}$ , job-finding probabilities  $\{m(\theta_p), p_g\}$ , stocks of inactive  $\{i_{p,m}, i_{p,f}, i_{g,m}, i_{g,f}\}$ , unemployed  $\{u_{p,m}, u_{p,f}, u_{g,m}, u_{g,f}\}$ , employed  $\{e_{p,m}, e_{p,f}, e_{g,m}, e_{g,f}\}$ , and private and public sector wages  $\{w_{p,f}, w_{g,f}, w_{p,m}, w_{g,m}\}$ , such that,  $\{\pi_m, \pi_f, \bar{e}_g\}$  and an exogenous “wedge” for female private sector wages  $\{\alpha\}$ :

- 1 Private sector firms satisfy the free-entry condition.
- 2 Male private sector wages are a constant fraction of workers' productivity.
- 3 Female private sector wages prior to applying a “wedge” are such that the value of a job for a firm is the same when hiring a man or a woman.
- 4 Newborns decide optimally which sector to join.
- 5 Workers decide optimally the threshold values of  $x$  for quitting their job or to stop searching.
- 6 Worker flows in and out of the four stocks are constant.
- 7 The total population adds up to 1 (0.5 men, 0.5, women):

- $\frac{1}{2}(1 - \Xi_m(\bar{e}_m)) = i_{g,m} + u_{g,m} + e_{g,m}^a + e_{g,m}^{na}$
- $\frac{1}{2}\Xi_m(\bar{e}_m) = i_{p,m} + u_{p,m} + e_{p,m}^a + e_{p,m}^{na}$
- $\frac{1}{2}(1 - \Xi_f(\bar{e}_f)) = i_{g,f} + u_{g,f} + e_{g,f}^a + e_{g,f}^{na}$
- $\frac{1}{2}\Xi_f(\bar{e}_f) = i_{p,f} + u_{p,f} + e_{p,f}^a + e_{p,f}^{na}$

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## Exits

$$i_{i,j}(\lambda F_j(\hat{x}_{i,j}) + \tau)$$

$$u_{i,j}(\lambda(1 - F_j(\hat{x}_{i,j})) + \tau + m(\theta_i))$$

$$e_{i,j}(\lambda(1 - F_j(\hat{x}_{i,j})) + \tau + \delta_i)$$

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## Entries

$$\lambda(1 - F_j(\hat{x}_{i,j}))[e_{i,j} + u_{i,j}] + \tau(1 - F_j(\hat{x}_{i,j}))$$

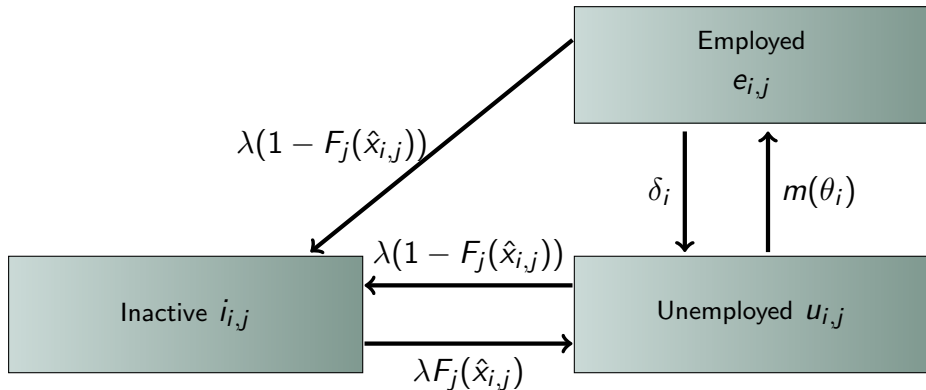
$$\delta_i e_{i,j} + \lambda F_j(\hat{x}_{i,j}) i_{i,j} + \tau F_j(\hat{x}_{i,j})$$

$$m(\theta_i) u_{i,j}$$

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← Return



Return

# Calibration - fixed parameters

	US (monthly)	UK (qt)	France (qt)	Spain (qt)
<b>Parameters set exogenously</b>				
<u>Discounting</u>				
Interest rate ( $r$ )	0.004	0.012	0.012	0.012
Death rate ( $\tau$ )	0.002	0.006	0.006	0.006
<u>Public sector policies</u>				
Wage premium (men) ( $\pi_m$ )	0.971	1.049	0.897	1.008
Wage premium (women) ( $\pi_f$ )	1.053	1.062	0.905	1.073
Employment ( $e_g$ )	0.120	0.170	0.137	0.074
<u>Labor market parameters</u>				
Matching efficiency ( $\zeta$ )	1	1	1	1
Matching elasticity ( $\eta$ )	0.5	0.5	0.5	0.5
<u>Time cost of labor force</u>				
Private ( $\xi_p$ )	1	1	1	1
Public ( $\xi_g$ )	0.971	0.964	0.915	0.944
<u>Arrival rate of shocks</u>				
Job separation - private ( $\delta_p$ )	0.018	0.019	0.021	0.055
Job separation - public ( $\delta_g$ )	0.009	0.007	0.008	0.027

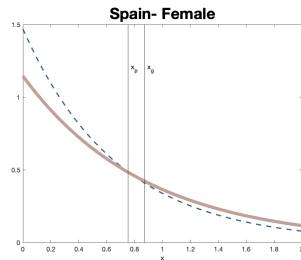
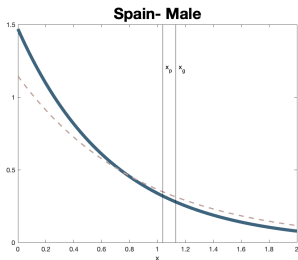
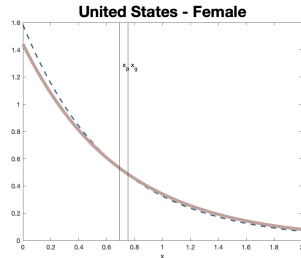
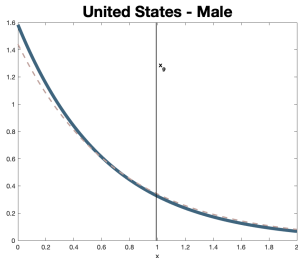
# Calibration - calibrated parameters

	US (monthly)	UK (qt)	France (qt)	Spain (qt)
<b>Calibrated parameters</b>				
<u>Labor market parameters</u>				
Bargaining power of men ( $\beta$ )	0.925	0.968	0.964	0.940
Cost of posting vacancies ( $\kappa$ )	3.158	0.914	1.398	1.411
"Wedge" female-male wage prv. sector ( $\alpha$ )	0.270	0.201	0.177	0.237
<u>Outside option distribution: Exponential</u>				
Mean - men ( $\mu_{x,m}$ )	0.632	0.570	0.743	0.681
Mean - women ( $\mu_{x,f}$ )	0.694	0.906	0.927	0.873
<u>Arrival rate of shocks</u>				
Outside option ( $\lambda$ )	0.081	0.079	0.065	0.106
<u>Preference distribution: Normal</u>				
Mean - men ( $\tilde{\epsilon}_m$ )	-85.792	-49.673	-9.959	-30.000
Mean -women ( $\tilde{\epsilon}_f$ )	-76.941	-24.377	-6.402	-25.326
Std. - men and women ( $\sigma_{\epsilon,m}$ )	84.670	51.889	13.598	21.910



# Calibration-distributions $F(\mu_x^j)$ , $j = [m, f]$

Return



Note: Means of these distributions for men (women) in each country are 0.632 (0.694) for the US and 0.681(0.873) for Spain. Thresholds for the public and private sector for men (women) are 0.990 (0.754) and 0.992 (0.695) in the US and 1.132 (0.872) and 1.039(0.754) for Spain.

# Calibration - targets

Targets	US		UK		France		Spain	
	Data	Model	Data	Model	Data	Model	Data	Model
Unemployment rate $((u_m + u_f)/((1 - i_m) + (1 - i_f)))$	0.064	0.064	0.060	0.060	0.093	0.093	0.168	0.168
Non-employment rates (full time equivalent)								
Male ( $i_m + u_m$ )	0.252	0.253	0.200	0.200	0.315	0.315	0.338	0.337
Female ( $i_f + u_f$ )	0.418	0.409	0.450	0.450	0.473	0.473	0.522	0.523
Private sector wage gap $w_f^p/w_m^p - 1$	-0.286	-0.289	-0.212	-0.212	-0.183	-0.183	-0.247	-0.248
Nr. of weekly wages- exp. cost vacancy $\kappa\Theta(1 - \eta)/(W_{mp}/4)$	8.000	7.997	8.000	8.000	8.000	8.000	8.000	7.998
Flows rates $E \rightarrow I$	0.022	0.022	0.020	0.020	0.021	0.012	0.032	0.032
Public sector employment shares ratio; full-time equivalent $(e_f^g/(e_f^p + e_f^g))/(e_m^g/(e_m^p + e_m^g))$	1.358	1.359	1.855	1.855	1.587	1.587	1.558	1.558
Ratio probability job finding private/public $p_g/m(\theta_p)$	1.066	1.066	0.743	0.743	0.809	0.809	0.878	0.878
Regional variation: Public sector size & over-representation $\epsilon_{u_g}/w_g$	0.004	0.004	0.002	0.002	0.007	0.007	0.011	0.011

◀ Return

# Drivers for women's over-representation: alternative decomposition

Country	Benchmark	No sector differences & no preference differences $\pi_j = 1$ $\xi_g = \xi_p$ $\bar{c}_f = \bar{c}_m$ $\delta_g = \delta_p$	Only wage differences $\xi_g = \xi_p$ $\bar{c}_f = \bar{c}_m$ $\delta_g = \delta_p$	Only hours differences $\pi_j = 1$ $\bar{c}_f = \bar{c}_m$ $\delta_g = \delta_p$	Only job security differences $\pi_j = 1$ $\xi_g = \xi_p$ $\bar{c}_f = \bar{c}_m$	Only sector differences $\bar{c}_f = \bar{c}_m$
Public-sector employment shares ratio						
US	1.359	0.991	1.171	0.993	0.985	1.166
UK	1.855	0.998	1.000	1.008	0.992	1.005
France	1.587	1.007	1.115	1.049	0.996	1.086
Spain	1.558	0.931	1.069	0.934	0.906	1.058

← Return

Percentage of wage private sector workers would sacrifice for public sector

$$PremiumH_i^p = \frac{(1 - \xi_g) \int_0^{\hat{x}_{p,j}} x f_j(x) dx}{F_j(\hat{x}_{p,j})} \frac{1}{w_{p,j}} \times 100$$

Country	Private sector worker [ $\xi_p = \xi_g$ ]		Public sector worker [ $\xi_g = \xi_p$ ]	
	Women	Men	Women	Men
US	0.84	0.81	0.87	0.83
UK	1.46	0.93	1.53	1.02
France	3.52	2.82	1.70	1.99
Spain	2.23	1.78	3.20	3.75

# Job-security premia

Consider a private sector worker, with

- wage  $w_1 = w_{p,j}$
- job-separation rate  $\delta_p$
- outside option  $x < \hat{x}_{p,j}$ .
- If offered a job-separation rate of  $\delta_g$ , what wage  $w_2$  would make him indifferent between the two options?
- $w_2 = w_1 + \delta_p(U_{p,j}(x|\delta_g) - E_{p,j}(x|\delta_p)) - \delta_g(U_{p,j}(x|\delta_g) - E_{p,j}(x|\delta_g))$
- where  $E_{p,j}(x|\delta_i) - U_{p,j}(x|\delta_i) = \frac{\xi_{p,j}(\hat{x}_{p,j} - x)}{r + \tau + \lambda + \delta_p + m(\theta_i)}$ .
- Integrating over  $x$  we calculate the conditional expected value:

Country	Private sector worker		Public sector worker	
	[ $\delta_p = \delta_g$ ]		[ $\delta_p = \delta_g$ ]	
	Women	Men	Women	Men
US	0.98	1.02	0.94	0.98
UK	0.92	1.17	1.20	1.52
France	1.43	1.66	1.70	1.99
Spain	2.88	3.39	3.20	3.75

Return

# Effects of public sector employment and wage policies

Policy	US	UK	France	Spain
<i>Increase of wages by 1 percent</i>				
$\Delta$ unemployment rate male	0.15pp.	0.11pp.	0.15pp.	0.07pp.
$\Delta$ unemployment rate female	0.27pp.	0.37pp.	0.31pp.	0.15pp.
$\Delta$ inactivity rate male	-0.13pp.	-0.11pp.	-0.12pp.	-0.06pp.
$\Delta$ inactivity rate female	-0.17pp.	-0.21pp.	-0.17pp.	-0.09pp.
$\Delta$ aggregate wage gap	-0.05pp.	-0.13pp.	-0.09pp.	-0.05pp.
<i>Increase of employment by 1 percent</i>				
$\Delta$ unemployment rate male	-0.07pp.	-0.09pp.	-0.05pp.	-0.03pp.
$\Delta$ unemployment rate female	-0.14pp.	-0.29pp.	-0.11pp.	-0.06pp.
$\Delta$ inactivity rate male	0.04pp.	0.04pp.	0.02pp.	0.01pp.
$\Delta$ inactivity rate female	0.04pp.	0.06pp.	0.02pp.	0.01pp.
$\Delta$ aggregate wage gap	-0.01pp.	-0.01pp.	-0.001pp.	-0.01pp.

Return

$$E_{i,j}(\hat{x}_{i,j}) = NE_{i,j}(\hat{x}_{i,j}). \quad (3)$$

$$\hat{x}_{i,j} = \frac{w_{i,j}}{\xi_i} + \frac{\lambda}{\xi_i}[A_{i,j} - B_{i,j}], \quad (4)$$

(5)

Establishes first link between wages and inactivity: lower wages  $\Rightarrow$  higher inactivity

# Value functions

We define two values functions for non-employed, one for unemployed and another one for inactive individuals.

$$\begin{aligned}(r + \tau + \lambda)E_{i,j} &= (1 - \xi_i)x + w_{i,j} + \delta_i[U_{i,j} - E_{i,j}] + \lambda[A_{i,j}^1 + A_{i,j}^2], \\(r + \tau + \lambda)U_{i,j} &= x + m(\theta_i)[E_{i,j} - U_{i,j}] + \lambda[B_{i,j}^1 + A_{i,j}^2], \quad \text{if } x \leq \hat{x}_{i,j} \\(r + \tau + \lambda)I_{i,j} &= x + \lambda[B_{i,j}^1 + A_{i,j}^2], \quad \text{if } x > \hat{x}_{i,j}\end{aligned}$$

where  $A_{i,j}^1 = \int_0^{\hat{x}_{i,j}} E_{i,j}(x')dF_j(x')$ ,  $A_{i,j}^2 = \int_{\hat{x}_{i,j}}^{\infty} I_{i,j}(x')dF_j(x')$ ,  
 $A_{i,j} = A_{i,j}^1 + A_{i,j}^2$ ,  $B_{i,j}^1 = \int_0^{\hat{x}_{i,j}} U_{i,j}(x')dF_j(x')$  and  $B_{i,j} = B_{i,j}^1 + A_{i,j}^2$ .

◀ Return