

Estimation of the gender pay gap in London and the UK: an econometric approach

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GLA Economics 17 May 2005

www.london.gov.uk/mayor/economic_unit



Overview

- One of the Mayor's priorities is to improve access to London jobs for all people.
- London's economic structure
- Economic activity rates of women
- Where are women in London working?
- Women on the board
- Women as business owners
- Gender pay gap





Introduction

This presentation will concentrate on two main issues:

1) Why we used STATA instead of other software



 2) How we used STATA to analyse the gender pay gap in London and the UK.
 Commands that we found useful for our research



Reasons for using STATA

- Version 8 is more user friendly than previous versions:
 - Choice of using menus for different commands and/or doing programming (DO FILES). More flexibility for programming than SPSS.
- Easy to handle large datasets and the facility of the matrix notation.
- Access to powerful statistical analysis, such as cross tabulations, percentiles in the distribution.
- Easy to run regressions for a cross sectional analysis or panel analysis for around 10,000 households with more than 40 independent variables.
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Our research interests

Understand differences between:

- Employment rates by gender in London compared to the rest of the UK.
- 2. Wages of men and women, in London and outside of London:
 - Wages for men and women by individual and job characteristics.
 - Matrix of wages by occupation and industry (for men and women).
 - Modelling the gender pay gap in London and the UK.





Data used and filtering

- Used Labour Force Survey (LFS):
 - 60,000 households in the UK and about 26,000 resident persons in Greater London.
 - Needed sample of those working age population by gender.
 Women working age is between 16-59 and for men is 16-64. Filtering data:

```
keep if (age >=16 & age <=64 & sex==1) | (age >=16 & age <=59 & sex==2)
```

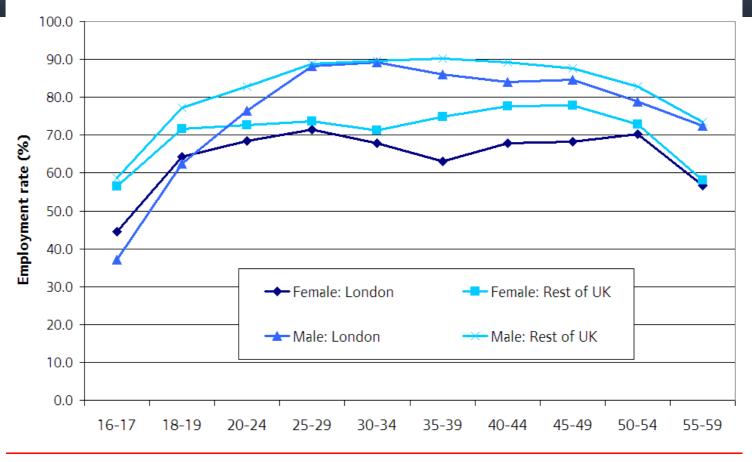
Provide two main broad applications using STATA in our research:

- 1) Calculation of employment rates for women and men and their wages by different individuals and job characteristics
- 2) Estimate unequal treatment between women and men:
 Decompose average gender pay gap in London into a)
 different individual and job characteristics and b)
 discrimination and/or unexplained factors.**GLA**ECONOMICS





Employment rates by gender and age



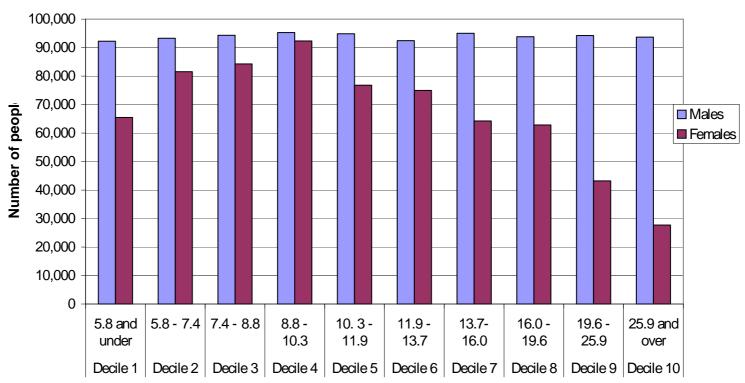
Employment = number of employees+ self employed + government employment/working age population
Source: Labour Force Survey (LFS), 2002/03

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Wages by deciles for men and women in London

Full time hourly earnings



Hourly pay (£) divided by male wage deciles

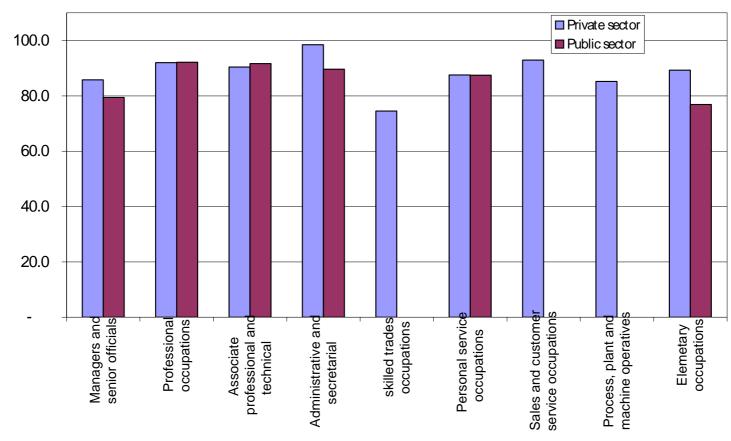
Source: LFS data, 2002/03





Gender pay gap by occupation and whether private or public sector

Female hourly pay as % of male hourly pay



Source: LFS data, 2002/03





Useful commands

- For standard cross tabulation: TABULATE variable
 1 variable 2 [fweight=name variable of weight] if all different conditions.
- To calculate wages at different points of the distribution with two different characteristics and count the number of individuals at that wage:
 TABLE characteristic 1 characteristic 2 (if all different conditions) [fweight=name variable of weight], contents (median wage count wage) row col





Estimating discrimination

- Implement Oaxaca's decomposition, which decomposes the mean gender pay gap into two main components:
 - I) Individual and job characteristics between men and women. Estimate wage differential in absence of discrimination.
 - II) Discrimination part (and/or unexplained factors):
 - A) Overpayment of men or
 - B) Underpayment of women





Oaxaca and Ransom's approach (1994)

Equation (1):

$$\ln \overline{W}_m - \ln \overline{W}_f = \overline{X}'_m (\hat{\beta}_m - \hat{\beta}^*) + \overline{X}'_f (\hat{\beta}^* - \hat{\beta}_f) + (\overline{X}_m - \overline{X}_f)' \hat{\beta}^*$$
(1A) (1B) (2)

where:

 $\ln \overline{W}_m$ and $\ln \overline{W}_f$, denote the mean logarithm wage for men and women respectively.



 \overline{X}'_m and \overline{X}'_f , are vectors of the mean values of the explanatory variables in the wage regression for men and women respectively.

 $\widehat{eta}_{\scriptscriptstyle m}$ and $\widehat{eta}_{\scriptscriptstyle f}$ are the vectors of estimated coefficients from the regression equations for men and women respectively.

 $\hat{\beta}^*$ is the estimated vector of coefficients from the equation of all workers.

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Estimation of mean gender pay gap

- Majority of studies estimate gender pay gap, considering how different male wages are from female wages (and two regression equations).
- To estimate the average gender pay gap, our research considered three regression equations:
 - 1. For all workers in London
 - 2. For all male workers in London
 - 3. For all female workers in London





Factors considered to explain pay gap

- Age (proxy of experience)
- Whether working public/private sector
- Whether working full-time/part-time
- Number of children
- Highest level of qualifications
- Ethnicity
- Industry
- Occupation
- Firm size
- Region





Commands used

1. regress Iwage age agesq public fulltime hdpch19 interactionchildpt degree highereducation gcea gcse otherqualification noqualification white mixed asian black chinese dce dd df dgh di djk dlmn dopq d1* d2* d3* d4* d521 d522 d523 d524 d531 d532 d541 d542 d543 d549 d6* d7* d8* d9* num_under25 num_more500 d_innerlondon [pw=pwt03] if govtof==8 matrix b_allwLondon = e(b) /*Obtained estimated coefficients of regression in a vector form for all workers

For men/women workers regression as above but used condition if male or female.

Get also matrix b_men = e(b) */estimated coefficients of regression for men

matrix b_women= e(b) */estimated coefficients of regression for women

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Commands used (2)

2. su Iwage age agesq public fulltime hdpch19 interactionchildpt degree highereducation gcea gcse otherqualification noqualification white mixed asian black chinese dce dd df dgh di djk dlmn dopq d1* d2* d3* d4* d521 d522 d523 d524 d531 d532 d541 d542 d543 d549 d6* d7* d8* d9* num_under25 num_more500 d_innerlondon [fw=pwt03] if govtof==8

For men/women workers (used condition if male or female). /*Obtained vector of means of each individual and job characteristics

3. Exported vectors of estimated coefficients and mean values of explanatory variables from the three regressions respectively into EXCEL. We calculated components (1A), (1B) and (2) from equation (1) to estimate the mean gender pay gap. This involves the multiplication of the vector of the mean values of the factors considered and the vector of estimated coefficients. In STATA we could see that the second secon





Key findings on gender pay gap

- Differences in individual and job characteristics account for most of the gender pay gap.
- The impact of direct unequal treatment appears to be slightly lower in London than outside.
- London part-time workers of both sexes are paid less than full-time workers.
- Among full-time workers, lower paid workers have virtually no difference in pay between men and women however the gender pay gap widens further up the wage distribution to 24 per cent for the top decile.
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Final thoughts

- We will continue to use STATA in our future research.
- We would like to calculate the Oaxaca decomposition within STATA using a more elegant programming solution.
- Our results from this study are published as Women in London's Economy and available at www.london.gov.uk/mayor/economic_unit
- Thank you for listening!

