The role of earnings expectations versus non-pecuniary factors in university attendance

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Sciences Po

- · Lots of evidence of benefits of a university degree
 - Increased earnings
 - · Better health

- · Lots of evidence of benefits of a university degree
- Often taxpayers are paying for these benefits
 - £17 billion upfront costs of higher education in the UK in 2017 (Dearden et al., 2017)
 - OECD countries' public spending on HE \approx 1% of GDP (OECD, 2020)

- · Lots of evidence of benefits of a university degree
- Often taxpayers are paying for these benefits
- Persistent gap in HE attainment by socio-economic status
 - · and hence in who enjoys these state-sponsored benefits
 - England: children of parents' earning in top 20% twice as likely to attend university as children of parents in the bottom 20%
 - "SES-gap" in attainment up to 30pp across OECD countries (OECD, 2018)
 - · is HE actually hindering social mobility?

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Understanding the factors that influence educational attainment is key, not only for educational outcomes but also for wider issues such as inequality and who benefits from public spending

- How important are earnings expectations vs other factors in decision to attend university?
- What's driving the SES-gap in educational attainment?
- How have these factors changed over a period of expansion in HE attainment (1980s to today)?

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 - US: Heckman et al. (2006) estimate earnings and "psychic costs"
 - France: D'Haultfoeuille and Maurel (2013) similar exercise
 - both rely on family background heterogeneity to estimate non-pecuniary factors
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 - 10% response rate, students still at university
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- Panel data for cohort born in 1989/90
- Model + data \rightarrow estimate distributions of factors
- Re-estimate model on cohort born in 1970

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 - · Extended-Roy model of educational choice at age 16 / 17
 - · Explicitly include earnings expectations and other factors
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- · Model the decision to go to university
- Panel data for cohort born in 1989/90
 - Schooling, background, and non-pec. expectations from before decision
 - Choices: did they complete university? (96% completion rate)
 - · Earnings and occupation after entry to labour market (age 25)
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 - · Compare across cohorts
 - \rightarrow understand drivers of expansion of HE in England

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What do students consider when making educational choices?

• (future) earnings

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 - + meet friends / partner
 - \pm studying
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- · life after university:
 - + better job / career
 - \pm graduate "identity"
 - debt

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- θ_S is a vector of expectations about other aspects of life
- *ϵ_S* is a random utility term
- terms differ for each individual conditional on choice S

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A model of educational choice

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• probability of attending university conditional on Y_s^{ea} and θ_{Si} is

$$\Pr(S = 1 | Y_s^{ea}, \theta_{Si}) = \Pr(\alpha(Y_{1i}^{ea} - Y_{0i}^{ea}) + (\theta_{1i} - \theta_{0i})'\gamma > \epsilon_{0i} - \epsilon_{1i})$$

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British Cohort Study (comparison over time)

- · Similar study also run by CLS
- Cohort born in one week in April 1970
- · Waves every 4 years since birth

Sample design

- Sampling "unit" school: 647 of 892 selected took part
- · Repr. sample oversampled minorities and "deprived" schools
- 15,770 interviews in sweep 1, from sample of 21,000 (74%)
- Sweeps annually between 14 and 20, then again at 25

- subjective: captures which aspects each student considers
- open-ended: students can mention anything, not leading q's
- · questions specifically about attending university
- · Similar responses identified and harmonised by survey designers
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Subsample: all students who were asked subjective questions about university (those with > 5 GCSEs)

	Full sample	Subsample
Ν	6,628	4,640
Female	0.55	0.57
Degree*	0.58	0.68
Russell group*†	0.26	0.28
Employed*	0.83	0.87
Wage (GBP)*‡	393	424

Notes: * At age 25. [†] Among degree holders. [‡]Median wage.

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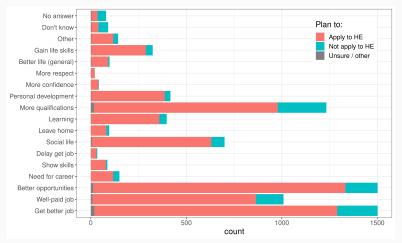
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Advantages of attending university

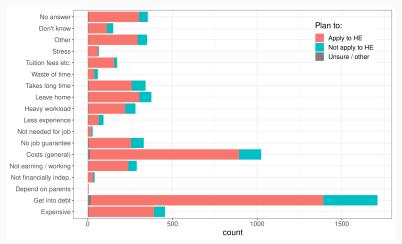
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What do you think the disadvantages, if any, might be for someone of going to university to study for a degree?

Disadvantages of attending university

What do you think the disadvantages, if any, might be for someone of going to university to study for a degree?



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where $\ensuremath{Y_S}$ are realised earnings at age 25 given choice \ensuremath{S} (wage distribution

Key assumptions

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 - · also include family background and other characteristics



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standard in discrete-choice models

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 - no. A-levels taking; gender; whether think high pay is important
- recall: θ contains responses to open-ended questions (plus components of *X*)

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1990 cohort By SES (parental income) Changes between cohorts

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· Compare relative importance of earnings versus other factors

- Compare relative importance of earnings versus other factors
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- Strategy:
 - Estimate parameters α , β_k , $\Delta\delta$

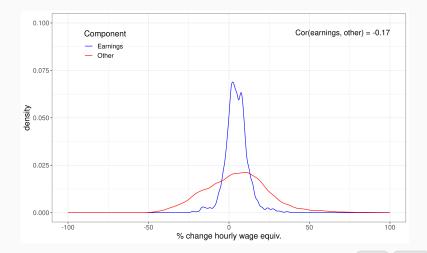
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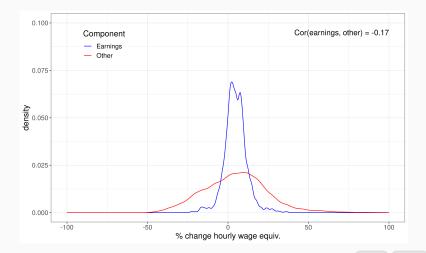
Full sample (1990): earnings vs other factors

• Both distributions similarly located, with positive means (4.8% vs 4.9%)



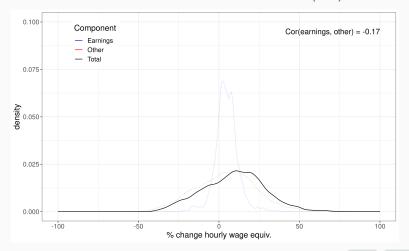
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 ⇒ Main determinant of decision is other factors: SD(total) = 19.2%



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 \rightarrow only 24.0% have high-enough earnings expectations to choose to attend university

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 - ightarrow now 99.5% attend university
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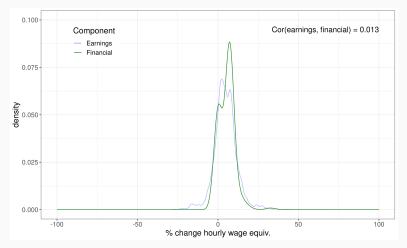
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Full sample (1990): splitting *t* into financial and other

· Financial factors distribution very similar to earnings expectations

- Mean financial positive (5.4%), mean other now negative (-2.1%)
- SD financial smaller (4.9%) than other (17.8%)

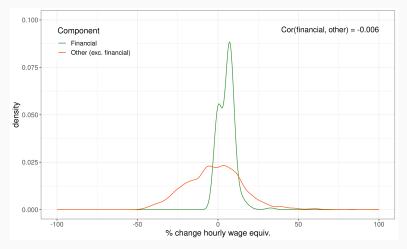
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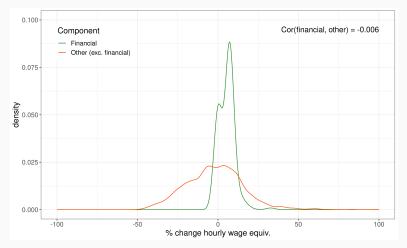
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- Use same parameter estimates as before
- Split sample by parental earnings at 16 into 3 groups:

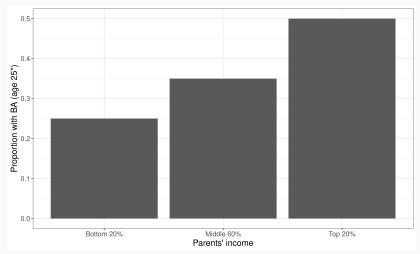
 \rightarrow bottom 20%, middle 60%, top 20%

• plot distribution of factors within each group

Why?

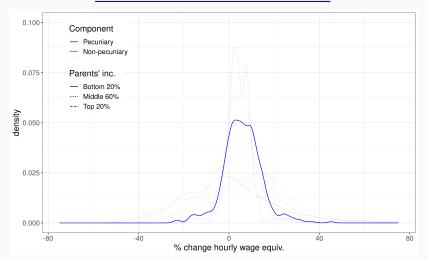
Huge SES-gap in attainment

Children of top 20% by income **twice as likely to hold BA at 25** than children of bottom 20%



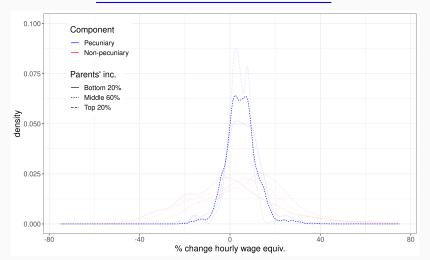
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Parental income	Mean	Variance	Skewness
Bottom 20% Middle 60%	0.147 0.143	0.018 0.013	0.335
Top 20%	0.099		0.650

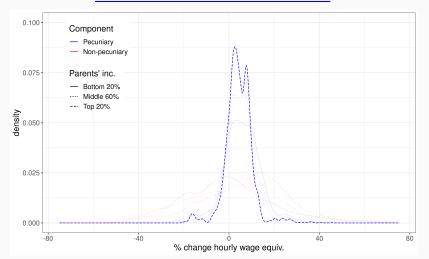


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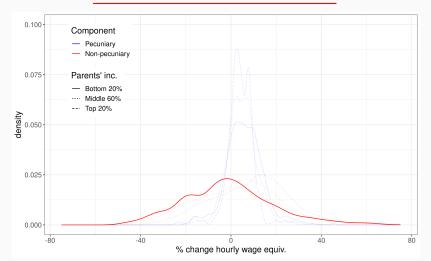
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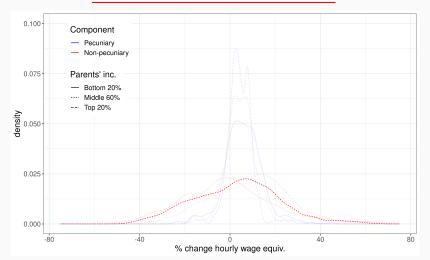


Parental income	Mean	Variance	Skewness
Bottom 20%	-0.185	0.066	0.520
Middle 60%	-0.183		0.524

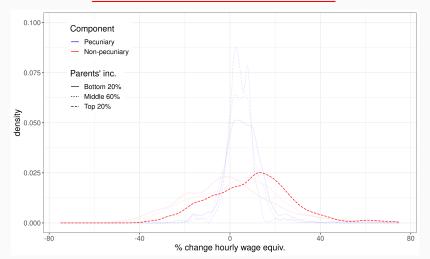


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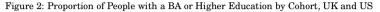


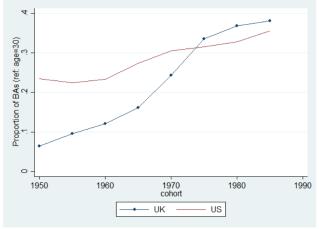
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27

Huge expansion in HE attainment



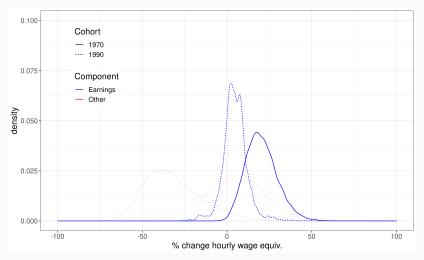


Note: sample restricted to age 22-59 and excludes full-time students. Each education-cohort cell has at least 100 observations.

- Re-estimate model on data from similar cohort study born in 1970
 - \Rightarrow Compare factors in 1970 to 1990

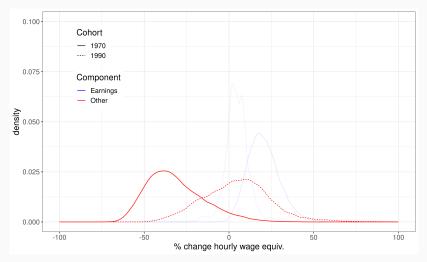
Changes between 1970 and 1990 cohorts

- earnings exp. fell: 20% \rightarrow 4.8% (mean); 9.3% \rightarrow 7.1% (sd)
- non-pec. benefits increased: $-31\% \rightarrow 4.9\%$ (mean); $17\% \rightarrow 20\%$ (sd) \Rightarrow non-pec. factors driving the large increase in attendance



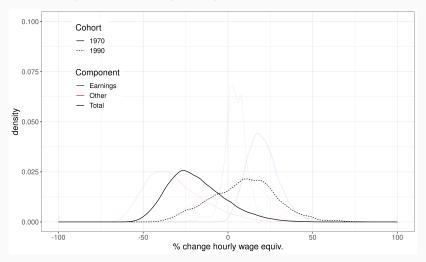
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Introduction

Extended Roy-model including psychic costs

Large-cohort panel data

Identification and estimation

Results

Conclusion

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- · Earnings exp. less important than other factors in HE decision
- · Earnings exp. similar across SES, other factors driving gap
- · Other factors drove large increase in HE attainment since 1970s

Still lots to do:

- 1. Continue to decompose other factors into meaningful components
 - Form natural "groups" of factors: Career; financial (now and future); Social life / environment; Education; Personal development; Time
 - But ind. parameters \rightarrow smaller groups difficult to interpret
- 2. Improve the model of earnings expectations
 - · Model lifetime earnings, rather than relying on single point
 - Allow for **unobserved heterogeneity** in earnings [see e.g. Heckman et al. (2006)]

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Thank you :)

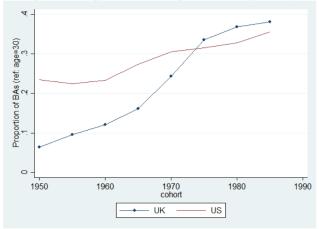
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UK HE Expansion

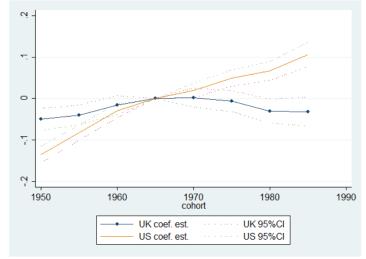
Figure 2: Proportion of People with a BA or Higher Education by Cohort, UK and US



Note: sample restricted to age 22-59 and excludes full-time students. Each education-cohort cell has at least 100 observations.

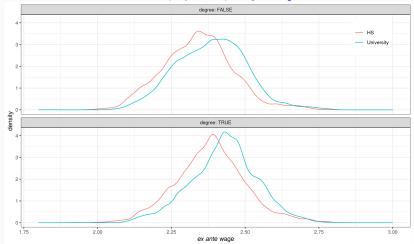
Graduate-wage premium

Figure 3: Ratio of BA median wage to that of high-school graduates, cohort effects



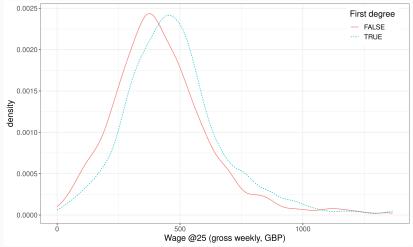
Wage distributions (Next Steps)

Ex ante (expected) wages, Y_s^{ea}



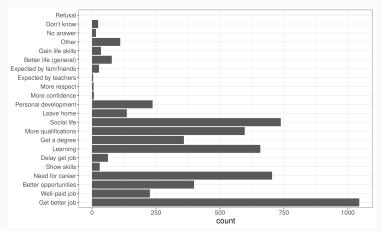
Wage distributions (Next Steps)

Ex post (realised) wages, Ys



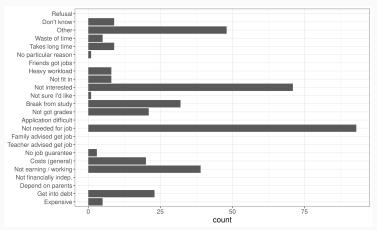
Main reasons for applying to university

Now thinking about yourself. You said you plan to apply for a place at university. What are YOUR main reasons for wanting to go to university?



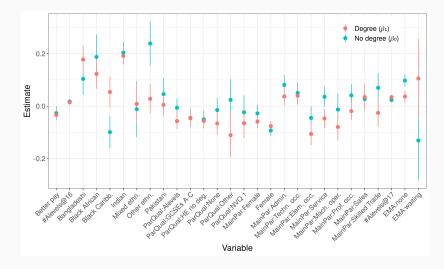
Notes: Students with >5 GCSEs @A*–C and who plan to apply. Open-ended. N = 4,640.

You have said that you are not planning to apply to university/not likely to ever apply to university. What are the MAIN reasons why you decided not to apply for a place at a university?



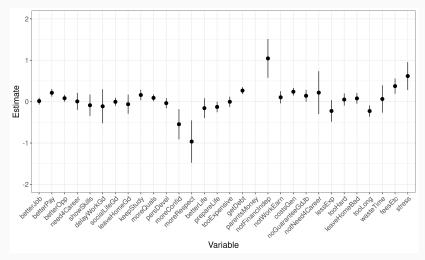
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Wage equation parameters, β_S



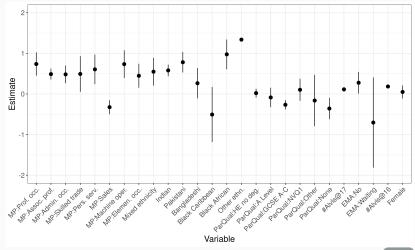
Choice equation parameters, γ

Responses to open-ended questions



Choice equation parameters, γ





back

Financial factors

Responses to open-ended questions classified as financial factors for decomposition of "psychic costs" (all *disadvantages*)

