To what Degree? - Recovering changes in the UK's graduate skill distribution.

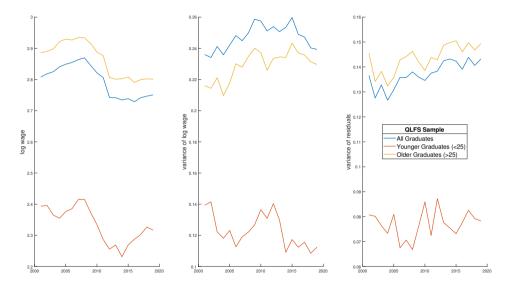
Max Schroeder

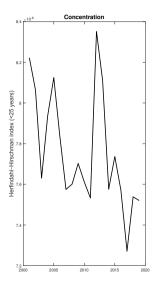
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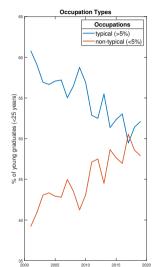
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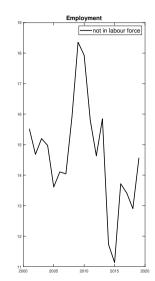
Introduction

- Generally, we believe a university education is beneficial for the individual, providing skills, good jobs and high earnings. But the HE sector is also crucial in determining the supply of high skilled labour allowing the economy to grow and develop.
- As a result, university education has become an increasingly important fixture of the UK education policy. Over the last 50 years, we have seen a massive increase in enrolment rates and a large increase in the number of HEI's.
- Recently, concerns have been raised about the effectiveness of universities in providing the correct skills to their students, leading to debates about inequality, employability, under-employment and cost-effectiveness.









- The last 20 years saw huge changes in both the demand for and supply of skills:
 - technological change
 - structural transformation of the labour market
 - higher education expansion
 - school reform
 - demographic change
- If we want to understand changes to the labour market outcomes of graduates, we need to take into account both supply and demand factors.

The qualitative margin

- Recent work suggests that subject of study matters a lot for graduate outcomes (e.g. Altonji et al. (2016), Andrews et al. (2022) and Lovenheim & Smith (2022)).
- Further work suggests that task content is an important driver of differences in labour market outcomes (c.f. Autor et al. (2003), Autor & Handel (2013), Sanders & Taber (2015))
- This suggests we should try and study the qualitative dimension of graduate's skills.

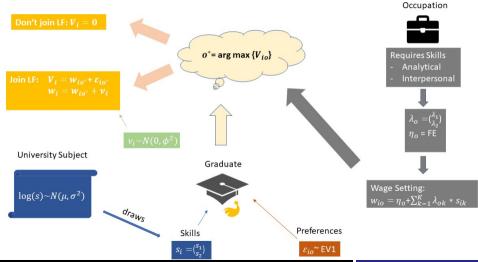
Aim of this paper

- In this paper I develop a model of the labour market for young university graduates. The model features:
 - multidimensional skills
 - skill heterogeneity
 - differences in the return to these skills
 - non-pecuniary preferences
- I estimate the model and use it to decompose the different factors driving changes to the labour market outcomes of young graduates.

The Economic Environment

- There is a finite number of Occupations, that differ in their demand for a finite number of general skills according to the tasks that are required to produce their output in each case.
- Similarly, workers are heterogeneous with respect to the multidimensional set of skills that they have.
- Workers choose which occupation they want to enter and workers are paid their marginal product Perivation

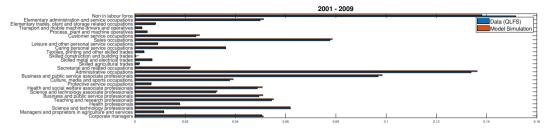
A picture says more than e^6.90775527898 words

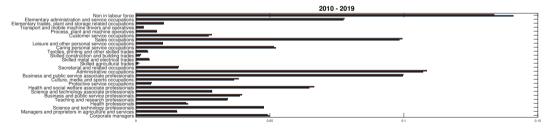


Taking the model to the data

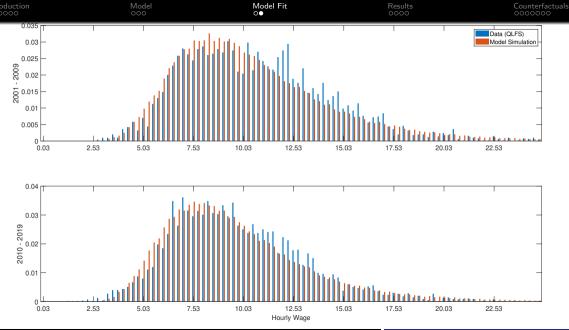
- Split sample into 2 periods: 2001 2009, 2010 2019. Sample details
- Aggregate degree subjects: Medical & Life Sciences, STEM, Business & Economics, Arts & Humanities, Other
- Two types of skill: abstract & interpersonal. Skill prices are estimated from SES. Skill Prices Task Examples
- Ask the computer (nicely) to maximize the LL given the observables while integrating out the distribution of unobservable skills.
 Estimation

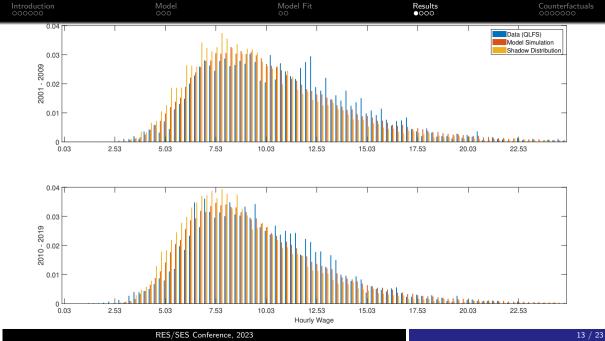


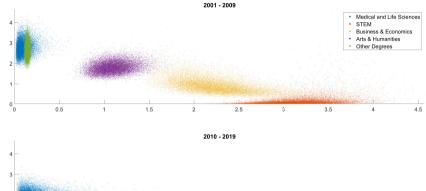




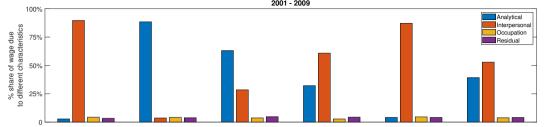
Source: QLFS and Model Simulation

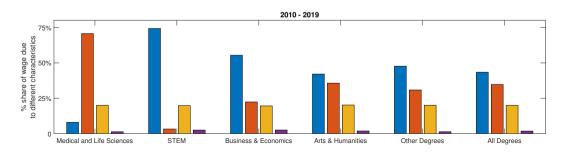




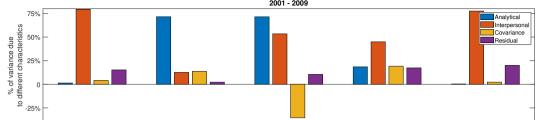


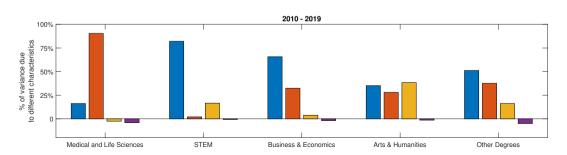






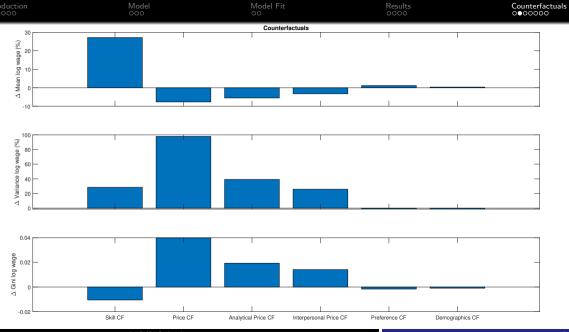


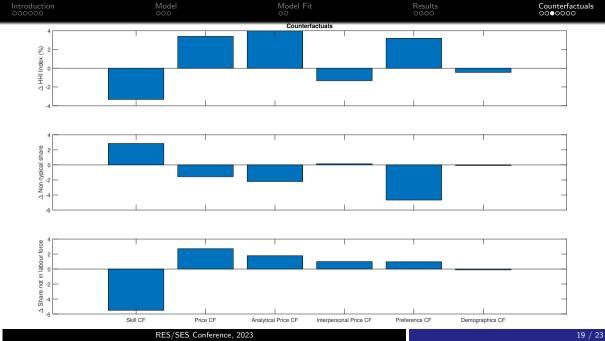




Counterfactuals

- Having estimated the parameters of the economy, I perform some counterfactual analyses by keeping different parts of the parameter space fixed at the period 1 values. Specifically consider:
 - Meeping the skill distribution fixed.
 - Keeping the skill prices fixed at 2001 values (abstract, interpersonal & both).
 - Seeping demographic composition fixed.
 - Meeping non-monetary preferences fixed.





Counterfactuals

- ullet Skill Counterfactual: higher mean wages, moderately higher variance (lower tail), less unemployment ullet Interpersonal skills matter for young graduates wages
- ② Price Counterfactual(s): small wage effect, large increase in variance (analytic > interpersonal), more concentration, less non-typical (analytical!) → Changing demand for analytical skill drew graduates into non-typical occupations.

Counterfactuals

- Oemographic Counterfactual: very little effect on either outcome
- ullet Preference Counterfactual: some impact on occupation concentration o non-typical occupations have become more attractive

Conclusion - what have we learned?

- The qualitative dimension of labour matters. Both for graduates and for the jobs that they go into. Just looking at high level descriptors (degree subjects, occupation classifications) obscures large changes within these categories.
- Over the last 20 years, analytical skills have become more important for graduates' labour market outcomes, driving a reduction in wage inequality and leading to more varied career choices.
- But reduction in interpersonal skills might be a cause for concern, especially if it impedes graduates' chance to advance into leadership and managerial positions.

The end.





Derivation of marginal product

$$h_{io}=e^{\sum_{K}\lambda_{ok}s_{ik}}$$

(Workers HC)

$$Y = F(H_1, ..., H_O)$$

(Aggregate Production Function)

$$\frac{\partial Y}{\partial h_{io}} = \frac{\partial F}{\partial H_o} \frac{\partial H_o}{\partial h_{io}} = \frac{\partial F}{\partial H_o} e^{\sum_K \lambda_{ok} s_{ik}}$$

(Take derivative)

$$\frac{\partial F}{\partial H_o} = \eta_o$$

$$w_{io} = \eta_o + \sum_{k} \lambda_{ok} s_{ik}$$

(Log Wage Equation)

▶ back



Data Sources

- Quarterly Labour Force Survey (2001 2019)
 - Degree Subject
 - Gross hourly wages (CPI deflated)
 - Occupation (2 Digit SOC)
 - Age
 - Gender
 - Geographic location
- Skills and Employment Survey (2001, 2006, 2012, 2017)

► Examples

Focus on fresh graduates (≤ 24 years). Resulting in a sample of 46,637 individuals.



Simulated Maximum Likelihood

• By combining the elements of the model, we can form the likelihood function:

Likelihood Function

$$\Pr(o_{i}^{*}, w_{i}^{obs}) = \int \left(\underbrace{\frac{e^{\eta_{o^{*}} + \sum_{K} \lambda_{o^{*}k} s_{ik}}}{\sum_{o=1}^{O} e^{\eta_{o} + \sum_{K} \lambda_{ok} s_{ik}}}}_{Occupation} \underbrace{\frac{e^{(-\frac{\nu_{i}^{2}}{2\phi^{2}})}}{\sqrt{2\pi\phi^{2}}}}_{Wage} \right) f(s)d(s)$$



- The estimation procedure works via *simulated maximum likelihood* we find the vector of economic parameters θ , that maximizes the probability of observing the actual outcomes.
- Use 1,000 quasi-random Halton draws to approximate the integral.
- Additional controls: Year and geography Fixed Effects, experience, part-time status and gender. As well as occupation specific non-pecuniary benefits.

- Find $Pr(o^*, w_i^{obs}|s_i; \theta)$ using the structure of the economic model.
- ② Integrate over the unobserved skill distribution $f(s|\theta)$: $Pr(o^*, w_i^{obs}|\theta)) = \int Pr(o^*, w_i^{obs}|s_i;\theta))f(s|\theta))ds$.
- $oldsymbol{\circ}$ Find vector of parameters $\hat{\theta}$ that maximizes the Log Likelihood function:

$$II(\theta) = \sum_{i=1}^{N} \log(\Pr_i(o^*, w_i^{obs}|\theta)))$$

▶ back

Derivation of Likelihood

$$\Pr(o_i^*, w_i^{obs}) = \int \Pr(o_i^*, w_i^{obs} | s_i) f(s) d(s) \qquad \text{(Rewrite unconditional probability)}$$

$$\frac{\Pr(o_i^*, w_i^{obs} | s_i)}{\Pr(o_i^* | s_i)} = \Pr(w_i^{obs} | s_i, o_i^*) \qquad \qquad \text{(Bayes Rule)}$$

$$\Pr(o_i^*, w_i^{obs} | s_i) = \Pr(o_i^* | s_i) * \Pr(w_i^{obs} | s_i, o_i^*) \qquad \qquad \text{(rearange)}$$

$$\Pr(o_i^*, w_i^{obs}) = \int \Pr(o_i^* | s_i) \Pr(w_i^{obs} | s_i, o_i^*) f(s) d(s) \qquad \qquad \text{(Integrate over skill distribution)}$$



Derivation of Likelihood cont.

$$\Pr(o_i^*|s_i) = \frac{e^{\eta_{o^*} + \sum_K \lambda_{o^*k} s_{ik}}}{\sum_{o=1}^O e^{\eta_o + \sum_K \lambda_{ok} s_{ik}}} \qquad \text{(Logit Choice Probability)}$$

$$\Pr(w_i^{obs}|s_i, o_i^*) = \frac{e^{\left(-\frac{\nu_i^2}{2\phi^2}\right)}}{\sqrt{2\pi\phi^2}} \qquad \text{(Normal PDF)}$$

$$\Pr(o_i^*, w_i^{obs}) = \int \left(\left(\frac{e^{\eta_{o^*} + \sum_K \lambda_{o^*k} s_{ik}}}{\sum_{o=1}^O e^{\eta_o + \sum_K \lambda_{ok} s_{ik}}} \right) \left(\frac{e^{\left(-\frac{\nu_i^2}{2\phi^2}\right)}}{\sqrt{2\pi\phi^2}} \right) \right) f(s)d(s) \qquad \text{(insert)}$$

▶ back



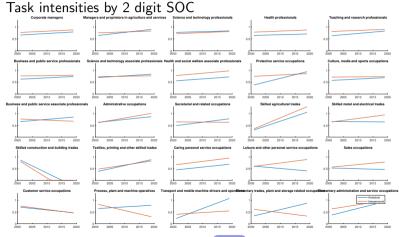
Tasks (with Examples)

- Abstract Tasks
 - Specialist knowledge and understanding
 - Complexity level of computer use
 - Advanced Mathematics/Statistics
- Interpersonal
 - Making Speeches/Presentations
 - Persuading or influencing others
 - Cooperating with colleagues





Skill Prices



Source: SES, authors own calculations back

