

# Socio-Economic Gaps in University Enrollment: The Role of Perceived Pecuniary and Non-Pecuniary Returns

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## Abstract

We elicit students' beliefs about different pecuniary and non-pecuniary benefits of university education in a sample of 2,540 secondary school students. Estimates of a dynamic choice model reveal that differences in perceived returns across socio-economic groups explain a substantial share of the socio-economic gap in intentions to enroll. Students who would be the first generation in their family to go to university perceive both the pecuniary and the non-pecuniary returns to university education as significantly lower. Among the non-pecuniary factors, beliefs about whether one would enjoy studying, perceptions about parental approval, and expected job satisfaction play the most important role.

JEL: I24, I26, J13, J24, J62

Keywords: Higher education, beliefs, socio-economic inequality, intergenerational mobility

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# 1 Introduction

Students from low socio-economic backgrounds are significantly less likely to attend university compared to students from more advantaged backgrounds with similar levels of prior academic achievement.<sup>1</sup> In the UK, students with at least one parent holding a university degree are about 14 percentage points more likely to go to university compared to students with similar levels of skills but less well educated parents.<sup>2</sup> The decision to attend university is a life-changing decision with a large impact on labor market, health, marriage, and crime outcomes.<sup>3</sup> It is therefore essential to understand why students with low socio-economic status are less likely to go to university. This open question is of high policy relevance given the low levels of socio-economic mobility in the UK and in many other countries where educational attainment and income are highly correlated across generations (Blanden, Gregg and Macmillan 2007; Black and Devereux 2011; Chetty et al. 2014).

While traditional models have emphasized the importance of credit constraints in explaining the socio-economic gap in enrollment (see Keane and Wolpin 2001; Carneiro and Heckman 2002; Gayle, Berridge and Davies 2002; Cunha et al. 2006; Belley and Lochner 2007; Lochner and Monge-Naranjo 2012), it is not well understood why we observe socio-economic differences in university attendance in countries in which grants and loans are available to students from disadvantaged backgrounds.<sup>4</sup> Traditional choice models based on rational expectations about discounted future income streams fail to generate the enrollment gaps observed in the data. Instead, the models need to rely

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<sup>1</sup>See, for example, Machin and Vignoles (2004); Chowdry et al. (2013) for the UK and Bailey and Dynarski (2011); Chetty et al. (2014) for the US.

<sup>2</sup>We use data from the British Household Panel Study (BHPS) and the UK Longitudinal Household Survey (UKLHS) to calculate the socio-economic gap in university attendance conditional on a range of cognitive and non-cognitive skills. The results of this analysis are reported in Table A.1 in the Appendix and are robust to the inclusion of cohort fixed effects.

<sup>3</sup>See, for example, Oreopoulos and Salvanes (2011), Oreopoulos and Petronjievic (2013), and Heckman, Humphries and Veramendi (2018) for evidence on the pecuniary and non-pecuniary benefits of university education.

<sup>4</sup>All students resident in the UK are eligible for student loans that cover tuition and maintenance irrespective of their socio-economic background (<https://www.gov.uk/student-finance>). Students only need to repay the loan if they find a job in which they earn above a certain threshold. Students from low-income households are also eligible for maintenance grants which do not need to be repaid.

on a residual catch-all-term generally referred to as ‘psychic cost’ or ‘consumption value’, which is allowed to vary across groups with different background characteristics.<sup>5</sup> Summarizing the results in the literature, Heckman, Lochner and Todd (2006) note: “*The evidence against strict income maximization is overwhelming. However, explanations based on psychic costs are intrinsically unsatisfactory. One can rationalize any economic choice data by an appeal to psychic costs* [p. 436]”. To better understand socio-economic gaps in university enrollment, it seems crucial to obtain a better understanding of what ‘psychic costs’ actually represent, and whether these costs vary systematically across socio-economic groups.

In this paper, we shed light on students’ motives to obtain university education and explore to what extent differences in beliefs can account for the socio-economic gap in enrollment. First, we elicit students’ beliefs about different pecuniary and non-pecuniary returns to university education in a sample of 2,540 secondary school students in England (ages 13-18), and we document how students’ beliefs about returns differ by socio-economic status (SES). Here we do not only focus on students’ beliefs about their likely labor market outcomes later in life but also on students’ perceptions about what their lives are likely to be like during the 3-4 years after they finish secondary school. We classify students as low or high SES depending on whether students would be the first generation in their family to attend university. Second, we estimate a dynamic choice model that allows for different sources of heterogeneity across socio-economic groups, and we investigate the relative importance of students’ beliefs in their decision to go to university. Finally, we investigate to what extent the socio-economic gap in university enrollment can be explained by differences in students’ beliefs about the different immediate and later-life returns to university education.

To investigate the role of beliefs in educational investment decisions, it is not possible to rely on choice data alone. Observed choices can be consistent with many different combinations of beliefs and preferences (Manski 2004). To overcome this identification

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<sup>5</sup>See, e.g., Carneiro, Hansen and Heckman (2003); Cunha, Heckman and Navarro (2005); Heckman, Lochner and Todd (2006); Cunha, Heckman and Navarro (2006); Cunha and Heckman (2007, 2008); Carneiro, Heckman and Vytlačil (2011).

problem, it is important to obtain direct measures of individual beliefs about returns.

For this purpose, we collect primary survey data and elicit individual intentions to attend university as well as beliefs about a range of different immediate and later-life benefits of university education that are of a pecuniary and non-pecuniary nature. More specifically, we ask students to imagine scenarios in which they attend or do not attend university during the 3-4 years after they finish secondary school. We then elicit their perceptions about different immediate outcomes that relate to their lives during these 3-4 years (e.g. enjoyment of social life, enjoyment of study/work, parental support, financial struggles). Moreover, we elicit their beliefs about different later-life outcomes (at age 30) that relate to their experiences in the labor market (e.g. earnings, probability of job enjoyment). To account for potential differences in students' beliefs about the probability of succeeding at university, we further elicit individual perceptions about the likelihood of graduating and obtaining high grades, and we present students with three scenarios when eliciting their beliefs about later-life outcomes (no university degree, university degree with low grades, university degree with high grades). Finally, we also ask students to state how likely they think it is they would have to work alongside their studies if they chose to go to university and what their preferred field of study would be.

We use this rich individual-level data to estimate a dynamic choice model in which students face the following sequential decisions. First, students decide whether to attend university or not and whether to work alongside their studies. Second, conditional on enrollment, students face the choice of whether to complete university or drop out. Third, students decide whether or not to work once they have completed their education. We allow for different sources of observed and unobserved heterogeneity across socio-economic groups and for heterogeneity across individuals in terms of perceived returns, gender, perceived ability and probability of obtaining high grades at university. We model students' decisions to work alongside their studies as a function of their financial situation, and allow different immediate non-pecuniary benefits and costs of university education to differ depending on whether the student decides to work alongside uni-

versity or not. We estimate the parameters of the model using simulated method of moments (SMM). We assess the model fit in different ways and find that we can closely match both targeted and not targeted moments in the data.

Our analyses reveal three main findings which contribute to our understanding of what drives socio-economic differences in university attendance. First, relative to high SES students, low SES students perceive both the immediate as well as the later-life benefits of university education as significantly lower. For example, while both low and high SES students believe they are more likely to enjoy studying than working and more likely to earn more money later in life if they go to university, these differences are markedly more pronounced for high SES students. Second, the estimates of our dynamic choice model reveal that perceptions about the non-pecuniary returns to university play an important role in students' enrollment decisions. Students' beliefs about their own ability are also found to be important. Third, we find that 25% of the socio-economic gap in students' intentions to go to university can be explained by differences in students' beliefs about returns. Among the non-pecuniary factors, students' beliefs about the likelihood that they would enjoy studying, perceptions about parental approval, and expected job satisfaction are most important in explaining the socio-economic gap.

Given the large socio-economic gaps in students' beliefs about the returns to university education, a natural question to ask is whether students are on average correct in their beliefs. Are the pecuniary and non-pecuniary returns to university education actually lower for first-generation students? As students self-select into university, we cannot provide a definite answer to this question, but we provide supplementary evidence on socio-economic differences in university earnings premia and students' actual experiences at university. The evidence suggests that returns to university education may indeed vary with socio-economic background. More research will be needed into understanding what may be driving these gaps in returns and which policies may be effective in narrowing socio-economic gaps in returns and enrollment.

Our study builds on and contributes to several strands of the literature. First, the

study relates to the growing literature which investigates the role of beliefs in human capital investment decisions. Previous work has mainly focused on the role of actual and perceived *pecuniary* returns in explaining educational attainment (e.g. Dominitz and Manski 1996, Jensen 2010, Abramitzky and Lavy 2014, Attanasio and Kaufmann 2014, Kaufmann 2014, Almas et al. 2016). To the best of our knowledge, we are the first to elicit students' beliefs about different immediate and later-life benefits of university education that are of a pecuniary and non-pecuniary nature and to use this data to estimate a dynamic model of schooling to understand students' motives for university attendance.<sup>6</sup> Indeed, we find that perceptions about immediate non-pecuniary benefits play a very important role for the decision to enroll in university as well as for explaining the socio-economic gap in enrollment.

Second, our paper relates to the literature documenting how additional services and amenities provided by universities influence enrollment. Jacob, McCall and Stange (2018) find that demand-side pressures have pushed US colleges to increase expenditures on consumption amenities, such as student activities, sports, and dormitories. The number of applications received by a specific university has been shown to increase after successful basketball and football seasons (Pope and Pope 2009) and after improvements in a quality-of-life ranking focusing on non-academic amenities (Alter and Reback 2014). While these studies provide indirect evidence that students value non-pecuniary attributes of university life, we directly measure beliefs about a wide range of different pecuniary and non-pecuniary benefits, allowing us to examine how they relate to choices and differ across socio-economic groups.

Third, our study relates to the literature which examines the role of beliefs about pecuniary and non-pecuniary benefits in students' choice of major (Montmarquette, Canings and Mahseredjian 2002; Arcidiacono 2004; Arcidiacono, Hotz and Kang 2012; Beffy, Fougere and Maurel 2012; Zafar 2013; Stinebrickner and Stinebrickner 2014;

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<sup>6</sup>Notable recent exceptions include Attanasio and Kaufmann (2017) who investigate the role of perceived marriage market returns and Belfield et al. (2019) who show that there is a strong positive association between perceived enjoyment of university and stated intentions to continue in higher education.

Wiswall and Zafar 2015*a,b*; Hastings et al. 2016; Wiswall and Zafar 2018), high-school track (Giustinelli 2016), and which specific university to attend (Delavande and Zafar forthcoming). In contrast to these studies, our analysis addresses the question of whether or not to go to university (i.e. the extensive margin), rather than which specific major, high-school track or university to choose. Given the large potential gains from a college degree (Heckman, Humphries and Veramendi 2018), the decision whether or not to enroll is crucial for many later-life outcomes as well as for social mobility.

This paper is organized as follows. Section 2 provides details on the survey design we use to elicit students' beliefs. Section 3 provides details on the sample and the survey data. Section 4 documents how students with different socio-economic backgrounds perceive the pecuniary and non-pecuniary returns to attending university. Section 5 presents the dynamic choice model and estimation approach. Section 6 presents the estimates of the dynamic choice model, while Section 7 provides the results of different policy simulations. Section 8 discusses how beliefs may be formed and how beliefs may relate to actual returns while Section 9 concludes.

## **2 Eliciting Student Beliefs**

To investigate the role of students' beliefs in their decision to go to university, we develop a novel survey tool to elicit students' intentions to attend university and their beliefs about the returns to university education. Section 2.1 explains the main features of our survey design, while Sections 2.2 and 2.3 present the methodology we use to elicit beliefs about different immediate and later-life benefits, respectively. The full list of questions can be found in Appendix B.

### **2.1 Survey Design**

We elicit students' intentions to go to university and their beliefs about the returns to university education in a sample of 2,540 secondary school students (ages 13-18).

We ask all questions *prospectively* to minimize potential bias that could arise due to ex-post rationalization. Similar to other studies (e.g. Bleemer and Zafar 2018), we use the self-reported likelihood of continuing in full-time education as our main outcome measure. More specifically, we ask students to state how likely they think it is that they will go to university if they get the required grades (0-100%). This survey methodology has the advantage that it allows students to express uncertainty about their choice. It also allows us to obtain a measure of intended choices that is not conflated by students' beliefs about whether they would get accepted. In the UK, 48% of a given cohort of students continues to higher education (Department for Education Statistical Fiscal Releases 2016).

We elicit students' beliefs about different immediate and later-life returns to university education using hypothetical investment scenarios. In addition, we elicit students' perceptions about the likelihood of obtaining the required grades to go to university, graduating conditional on enrolling, and obtaining high grades conditional on graduating. Finally, we elicit their beliefs about how likely it is they would work alongside their studies if they decided to go to university. We use students' responses to these questions when estimating the dynamic choice model, as described in detail in Section 5.

We validate our survey tool in multiple ways. First, we re-survey a subsample of participants two months after the initial survey. This allows us to compute test-retest correlations to assess the reliability of our main outcome measure. Second, for final year students, we compare stated intentions to continue in full-time education against actual decisions to apply to university. In the UK, all students who wish to go to university need to apply through a centralized application system (Universities & College Admissions Service – UCAS) by a specific date. We re-survey individuals *after* the application deadline, which allows us to obtain information on whether students actually applied to university or not, and if yes, which subject field they chose to apply to. Third, we assess the reliability of responses by examining how students respond to questions that are similar but reverse-coded. Finally, we merge our survey data with administrative data on the schools the students currently attend, which allows us to



assess whether the mean responses of students within a given school are consistent with information at the school level.

We took great care to emphasize that the survey is completely anonymous. We did not collect any personal information such as names or addresses, and it was made clear to the students that neither the researchers nor the students' teachers can identify any individual respondent. We matched students across survey waves using a survey ID.

## 2.2 Beliefs about Immediate Outcomes

To elicit student beliefs about different immediate benefits and costs of higher education, we ask students to imagine what their lives are likely to be like during the 3-4 years after they complete secondary school (i) if they go to university and enroll in the subject field of their choice and (ii) if they do not go to university but start working instead. A typical undergraduate degree in the UK takes 3-4 years to be completed. We deliberately chose to make it explicit that the alternative is to start working because we did not want students to think about the possibility of doing a gap year before starting in higher education. We use probabilistic questions to elicit student beliefs about 13 different immediate outcomes that are of a pecuniary and non-pecuniary nature (see Table 1). The use of the probabilistic scale has the advantage that the responses are interpersonally comparable and more informative than responses on a Likert-scale (see Manski 2004). For example, we ask students the following two questions:

1. *If you go to university, how likely do you think it is that you will meet people with whom you easily get along with? [0-100%]*
2. *If you start working, how likely do you think it is that you will meet people with whom you easily get along with? [0-100%]*

As is standard in the literature, the subjective probability questions are preceded by a section which explains the use of the 0-100% chance scale and asks respondents to

answer some warm-up questions.<sup>7</sup>

The set of questions we ask relate to different aspects of the students' lives after they finish secondary school. To reduce the dimensionality and measurement error, and deal with the correlation between students' responses to conceptually related questions, we group the 13 stated likelihoods into six categories:

1. *Social life*: enjoy social life and activities; meet people you easily get along with; lose contact with family and friends; feel lonely
2. *Subject interest*: find material/work tasks interesting; enjoy studying/work
3. *Stress*: find material hard/workload high; feel stressed
4. *Parental support*: have parental support in choice
5. *Life partner*: find life partner
6. *Financial struggle*: struggle financially; have enough money; have financial support from parents

For all 13 survey items which relate to students' experiences during the 3-4 years after finishing school, we first calculate individual differences in stated likelihoods across the two scenarios (university minus work). When several responses are aggregated into a single category, we extract the first factor from the differences in responses. The resulting factors have a mean of zero and a standard deviation of one. For comparability, we also standardize the other differences in responses which only correspond to one survey item (parental support in choice and finding a life partner).

Before we present students with these questions, we ask them to report which field of study they would be most likely to choose if they decided to go to university. More specifically, students are able to choose between five different subject fields: (i) Arts and Humanities (e.g. languages, history, music, architecture, philosophy), (ii) Life Sciences

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<sup>7</sup>In order to familiarize students with the nature of probabilistic questions, we ask: '*What do you think is the percent chance that it will rain tomorrow?*'

(e.g. biology, medicine, pharmacy, psychology), (iii) Physical Sciences and Engineering (e.g. mathematics, computer science, physics, engineering), (iv) Social Sciences (e.g. economics, law, business) and (v) Education. In the scenario in which we ask students to imagine that they attend university, we explicitly ask them to imagine that they enroll in the degree of their choice.

Table 1: Overview of belief elicitation questions

Scenarios	Outcomes
<i>Immediate Outcomes</i>	
(1) Attend university	Enjoy social life and activities (0-100%)
(2) Do not attend university	Meet people you easily get along with (0-100%)
	Lose contact with family and friends (0-100%)
	Feel lonely (0-100%)
	Find material/work tasks interesting (0-100%)
	Enjoy studying/work (0-100%)
	Find material hard/workload high(0-100%)
	Feel stressed (0-100%)
	Have parental support in choice (0-100%)
	Find life partner (0-100%)
	Struggle financially (0-100%)
	Have enough money (0-100%)
	Have financial support from parents (0-100%)
<i>Later-life Outcomes</i>	
(1) University degree with high grades	Earnings (conditional on working)
(2) University degree with low grades	Have job (0-100%)
(3) No university degree	Enjoy job (0-100%)

Notes: Students are asked 26 questions regarding potential outcomes during the 3-4 years after finishing secondary school ( $2 \times 13$ ) and a total of 9 questions regarding potential outcomes at age 30 ( $3 \times 3$ ).

### 2.3 Beliefs about Later-life Outcomes

To elicit individual beliefs about the returns to university education on different later-life outcomes, we present students with three different scenarios: (i) going to university and graduating with high grades (First-class honors or Upper Second-class honors), (ii) going to university and graduating with low grades (Lower Second-class honors or Third-class honors), and (iii) not going to university, which we treat the same as going to university but not graduating. For each of these three scenarios, we ask students how likely they think it is they will have a job at age 30, enjoy the job they will be doing,

and what their likely earnings will be (conditional on working full-time and assuming no inflation).

To obtain a better understanding of how students think about their likely future performance, we also ask students how likely they think it is they will graduate if they go to university and how likely they think it is they will obtain high grades (First-class honors or Upper Second-class honors) conditional on graduating. Unlike in the US, dropout rates in the UK are low with only 10.6% of all students who start a degree failing to qualify.<sup>8</sup> There is, however, heterogeneity in how well students do at university. Conditional on starting a degree and graduating, 73% of all student obtain First-class or Upper Second-class honors, while 27% obtain Lower Second-class honors or Third-class honors. Our questionnaire design allows us to capture heterogeneity across respondents regarding how well students think they will do. This seems especially relevant in our context as students from different socio-economic backgrounds may hold different beliefs about their likely future performance.

Potential caveats of our study include that we do not elicit the perceived growth or variance of earnings conditional on full-time work. We do, however, capture important sources of uncertainty by allowing for differences in beliefs about the probability of completing university, obtaining high grades, and having a full-time job. Moreover, as explained in Section 5, utility from earnings is assumed to be concave in earnings allowing for risk aversion and diminishing marginal utility. Another simplification is that in terms of outcomes at age 30, we treat the scenario of dropping out of university the same as not attending. We pooled these two categories because returns to college attendance for students who do not graduate have been shown to be small and roughly offset by the costs in terms of forgone earnings (Hendricks and Leukhina 2018).

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<sup>8</sup>The percentage of full-time first-degree students who do not continue in higher education beyond their first year is 6.4%, while the percentage who do not obtain a degree at the end of their studies is 10.6%. We also note that 3.8% of all full-time first-degree students switch to and complete another degree at the same higher education institution and that 5.4% transfer to and complete a degree at a different higher education institution.

## 3 Data

In order to examine which beliefs are important for students' decisions to enroll in higher education and to gain a better understanding of what drives the socio-economic gap in enrollment, we collect primary survey data from secondary school students in England. In the following, we describe the characteristics of our sample (Section 3.1) and assess the validity of responses (Section 3.2).

### 3.1 The Sample

The study sample consists of 2,540 students aged 13-18 from 37 schools in England who are at a critical age as they are about to make the decision of whether or not to go to university. We collected the data using an online survey, which was distributed via the student mailing lists of schools that agreed to participate in the research study (see map in Appendix Figure A.1).<sup>9</sup> Approximately 10% of all students who were contacted to participate in the study chose to participate. The main survey data was collected in November 2016 (wave 1). A short follow-up survey was administered in January 2017 (wave 2), for which 319 of the 2,540 students participated. Students were incentivized to participate in the surveys through a prize draw of a voucher worth £100. The median time students needed to complete the survey was 13 minutes. 37% of the students in our sample are male with the average student being in Year 11 (see Table A.2).<sup>10</sup> 759 students in our sample are in their final year of secondary school education (Year 13).

While socio-economic status or social standing in society is based on a combination

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<sup>9</sup>We did not use any specific selection criteria to select the schools we contacted. The Department for Education provides lists of all secondary schools and sixth form colleges in England. We used these lists of potential schools and contacted the head teachers of a random subset of these schools in no specific order.

<sup>10</sup>Out of the 37 schools that participated, 6 are all-girls and 2 are all-boys schools. 45% of all respondents from mixed schools are male. Gender is balanced across socio-economic groups; among low SES students, the fraction of male respondents is 36.4% while it is 36.9% among high SES students (p-value=0.79). On average, both female and male students are in Year 11, which is also true for low and high SES students.

of different variables such as occupation, education and income, we focus on one particular facet of socio-economic status, namely parental education. We define students as high SES if at least one of their parents went to university and as low SES if they would be the first generation in their family to go to university. 55% of students in our sample report that at least one of their parents obtained university education. To shed more light on whether socio-economic differences in beliefs may be related to differences in access to information, we ask students about the number of people they know whom they could ask about university life and whether they have a sibling or older friend who has been to university. On average, low SES students know 4.5 people whom they can ask about their experiences at university, while high SES students report knowing 7.2 people (p-value<0.001).<sup>11</sup> 54% of low SES students and 68% of high SES students report having a sibling or older friend who has attended university (p-value<0.001).

Compared to the average school in England, the schools in our sample have a lower proportion of students who are eligible for free school meals (23% vs. 29%) and they send a higher proportion of students to university (67% vs. 48%). Consistent with those patterns, a higher proportion of students in our sample has at least one parent with a university degree relative to what would be expected in a nationally representative sample (55% vs. 40%).<sup>12</sup> Figure A.2 in the Appendix compares the distribution of GCSE English Literature and Mathematics grades in our sample to (i) the distribution of grades within the sampled schools and (ii) the distribution of grades in England. The students in our sample have higher grades compared to the average student in the sampled schools as well as the average student in England. While more research will be needed into how our results generalize to representative samples, we note that we oversample those students whom we are most interested in: the high-achieving students for whom going to university is actually a realistic option.

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<sup>11</sup>All p-values reported in parentheses are for two-sided t-tests testing differences in means across socio-economic groups.

<sup>12</sup>Source: Family Resources Survey.

## 3.2 Survey Validation

In order to assess the reliability of survey responses to the question how likely students think it is they will go to university if they get the grades, we re-survey a subsample of all students two months after the initial survey. Whilst it is possible that beliefs can change between the two survey waves, we would not expect major shifts in beliefs for most respondents. The Spearman rank correlation between individual beliefs about the perceived probability of going to university stated in waves 1 and 2 for those students who were not in their final year is 0.532 (N=202), which is comparable to the test-retest correlation other survey studies find for individual survey items (e.g. Falk et al. 2016).

As explained in Section 2.1, students in the UK who wish to go to university need to apply via a centralized system by a specific date. Given that we re-survey students after this deadline, we can examine whether the stated intentions of final year students correlate with their actual decisions to apply. Panel A of Figure 1 shows the mean perceived likelihood of going to university stated in the initial survey for final year students who chose to apply to university and final year students who did not apply to university (N=117). We can see that stated intentions correlate highly with actual choices. For students who chose to apply to university, the average stated probability in wave 1 is 93% compared to 28% for students who did not apply. The two means are statistically different from each other at the 1% level. Consistent with those results, Appendix Figure A.3 shows that the fraction of students who applied is increasing in students' stated intentions to apply. For students who were in their final year, we included an additional question to the second wave in which we ask students, who did not apply, about their perceived likelihood of applying in the future. The correlation between stated intentions to apply to university in wave 1 and stated intentions to apply to university in the future (for final year students who did not apply) is 0.82, which suggests that those students who did not carry out their plans yet are very likely to do so in the future.<sup>13</sup>

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<sup>13</sup>While attrition between the surveys is non-random, with students stating a higher likelihood of going to university being more likely to fill out the follow-up survey (82.6% vs. 77.8%), we find similar

We also investigate students' stated subject choices and how those relate to actual application decisions. In wave 1, 27% of all students stated that they would study Arts and Humanities, 27% Life Sciences, 21% Physical Sciences and Engineering, 20% Social Sciences and 6% Education.<sup>14</sup> Among the students who applied, 90% did indeed apply to a subject in the subject field they stated was their most preferred in wave 1. As can be seen in Appendix Figure A.4, low SES students intend to study similar subject fields as high SES students.

We additionally assess the reliability of our survey tool by investigating the Spearman rank correlations between different survey items, some of which are reverse-coded (see Appendix Tables A.3 and A.4). As expected, we do find that similar survey items correlate positively (e.g. enjoy social life and meet people with whom you easily get along with), while similar reverse-coded survey items correlate negatively (e.g. enjoy social life and feel lonely). We have illustrated some of the joint distributions of survey responses in Appendix Figure A.5. Similar patterns can be found for low and high SES students.

To further investigate the reliability of our response measures, we investigate whether students' mean responses within a given school correlate with school-level information we obtain from administrative data (see Appendix Figure A.6). The following findings emerge. First, students who attend schools in which a high proportion of students continue to higher education are more likely to perceive the likelihood of attending university to be higher. Second, students who attend schools with high results on national standardized tests (measured by the average Attainment 8 score, which is based on the students' GCSE results in Year 11) perceive the probability of obtaining the grades to go to university to be higher. Overall, we conclude that these patterns are consistent which strengthens our confidence in the reliability of the survey data.

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results in terms of validity of the survey measure for both socio-economic groups. More specifically, the test-retest correlation is 0.503 for low SES students and 0.478 for high SES students, and for both groups the stated intentions significantly differ by actual application decisions (at the 1% level).

<sup>14</sup>These numbers are comparable to statistics from the Higher Education Statistics Authority. Among all students, 22% study Arts & Humanities, 27% study Life Sciences, 20% study Physical Sciences and Engineering, 24% study Social Sciences and 7% study Education.



## 4 Socio-economic Differences in Beliefs

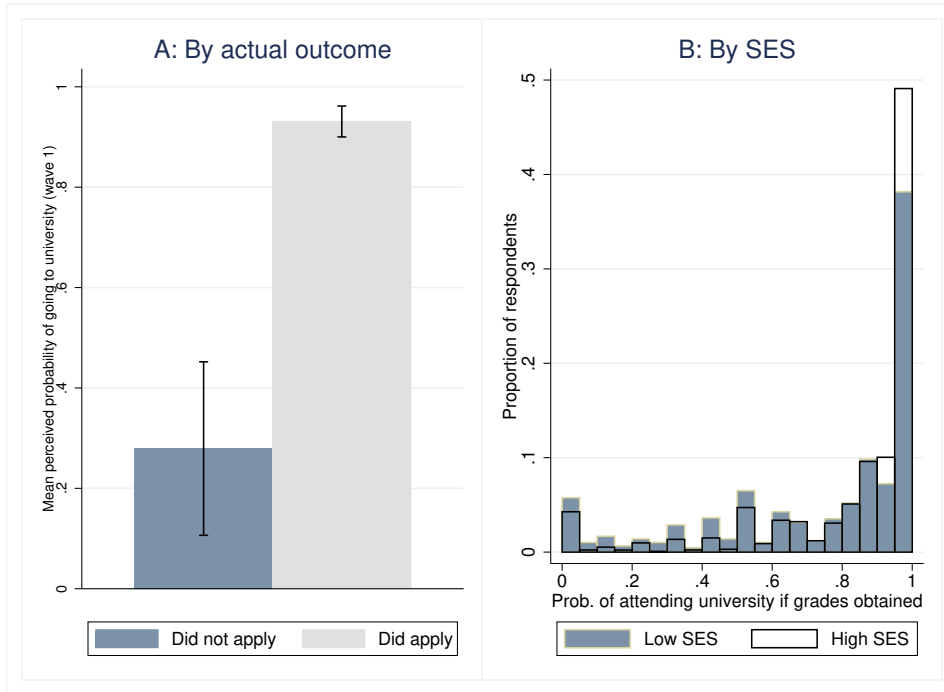
In this section, we investigate how students from different socio-economic groups differ in their intentions to go to university as well as in their beliefs about the future. We first examine students' responses to our main outcome measure, namely the perceived likelihood of going to university conditional on achieving the required entry grades (see Table 2). Panel B of Figure 1 shows the distribution of responses, separately for low and high socio-economic status individuals. The mean perceived probability of going to university is 83% for students who have at least one parent who attended university and 74% for students whose parents have not attended university (p-value<0.001). Whilst the figure shows a significant proportion of students in our sample who are virtually certain that they want to go to university (46% for high SES and 36% for low SES students), there are also a non-trivial proportion of respondents who are either not likely or deem it more unlikely than likely that they will go to university, thereby showing a substantial degree of heterogeneity in beliefs in our sample.

Table 2: Differences in intentions and perceived performance

Variable	All	SES		Difference (p-value)
		Low SES	High SES	
Attend university	0.784 (0.291)	0.741 (0.311)	0.830 (0.262)	0.089*** (0.000)
Achieve required grades	0.699 (0.222)	0.671 (0.220)	0.734 (0.211)	0.062*** (0.000)
Graduate from university	0.826 (0.207)	0.815 (0.219)	0.849 (0.182)	0.034*** (0.000)
Obtain high grades	0.672 (0.205)	0.646 (0.209)	0.705 (0.190)	0.058*** (0.000)
Work alongside studies	0.660 (0.277)	0.706 (0.263)	0.632 (0.283)	-0.074*** (0.000)

*Notes.* Column 1 presents the mean across all respondents while columns 2 and 3 show the means for low SES and high SES students, respectively. Column 4 displays the difference in means along with the corresponding p-value testing differences in means. High SES is defined as having at least one parent who has a university degree. Attend university: Stated likelihood of attending university if the required grades are obtained. Achieve required grades: Stated likelihood of obtaining the required A-level grades to go to university. Graduate from university: Stated likelihood of graduating from university conditional on enrolling. Obtain high grades: Stated likelihood of obtaining a First or 2.1 conditional on graduating. Work alongside studies: Stated likelihood of having to work alongside if enrolled in university. \* p<0.10, \*\* p<0.05, \*\*\* p<0.01.

Figure 1: Perceived probability of going to university



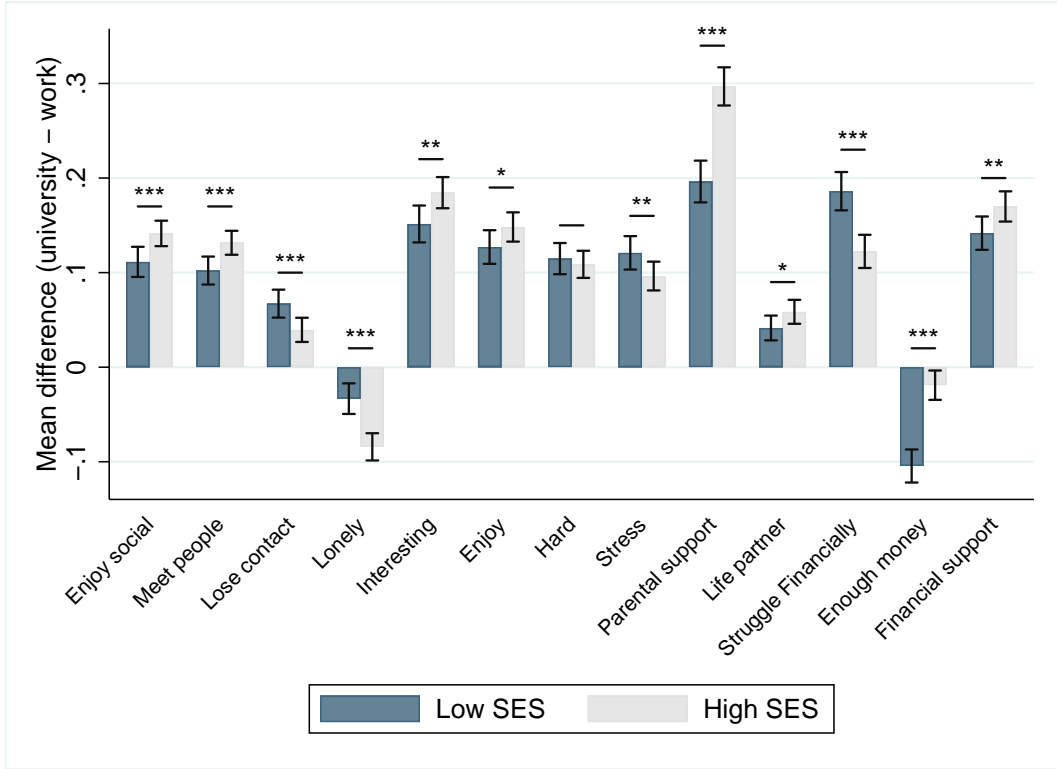
Notes: Panel A shows the probability of going to university stated in wave 1, separately averaged across individuals in their final year who applied to university and those who did not (with 95% confidence intervals). Panel B shows the distribution of stated beliefs of attending university separately for low and high SES students. High SES students are defined as those students who have at least one parent with university education.

As can be seen in Table 2, students from low SES backgrounds think it is less likely they will obtain the required grades to go to university (67% vs. 73%,  $p$ -value  $< 0.001$ ), graduate if they enroll (82% vs. 85%,  $p$ -value  $< 0.001$ ), and obtain high grades if they graduate (65% vs. 71%,  $p$ -value  $< 0.001$ ). At the same time, they state a higher likelihood of having to work alongside their studies if they go to university (71% vs. 63%,  $p$ -value  $< 0.001$ ).

Turning to socio-economic differences in beliefs about returns, low SES students perceive both the immediate and later-life returns to attending university as lower. Figure 2 shows average perceived immediate returns to attending university by SES (see Table 1 for a full list of questions).<sup>15</sup> Several notable findings emerge. On average, students in both socio-economic groups believe it is more likely they will enjoy their

<sup>15</sup>Appendix Table A.5 presents average beliefs about each immediate outcome in each of the two scenarios as well as average beliefs about returns separately for low and high SES respondents.

Figure 2: Perceived immediate returns by SES



Notes: The figure shows average differences in beliefs (university-work) regarding the immediate outcomes separately by socio-economic status. High SES students are defined as those students who have at least one parent with university education. Stars indicate differences by SES: \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

social life and meet people they easily get along with if they go to university rather than start working instead. They also think that it is less likely they will feel lonely but also more likely they will lose contact to their family/friends. Notice that there are significant differences across the two socio-economic groups in terms of how large these perceived immediate non-pecuniary benefits/costs are. In particular, students with low SES backgrounds perceive the return in terms of the likelihood of enjoying their social life (p-value=0.004) and meeting people (p-value=0.003) to be significantly lower, while they perceive the cost in terms of feeling lonely (p-value<0.001) and losing contact to their family/friends (p-value=0.005) to be higher.

When it comes to the tasks and workload associated with the different choices, we find that both groups on average expect the material/work tasks to be more interesting

if they go to university. They also think they would enjoy studying more than working. At the same time, they believe it is more likely they will find the material hard or the workload too high and that they will feel stressed. Students with low SES perceive the benefit in terms of how interesting (p-value=0.011) or enjoyable (p-value=0.077) the tasks are to be lower, but there is no SES gap in terms of finding the material too hard and there is only a small perceived gap in terms of the cost of feeling stressed. Moreover, we find large differences across socio-economic groups in terms of the perceived difference in parental approval. Both groups believe that parents will approve of their choice more if they go to university, but this difference is 10 percentage points larger for high SES students (p-value<0.001). Both groups perceive it to be somewhat more likely to meet their life partner if they go to university. The difference is again greater for high SES students (p-value=0.068).

In terms of financial factors, both groups of students are more likely to think they will struggle financially if they go to university and less likely to have enough money to do what they enjoy. They also think they will be more likely to be supported financially by their parents. Again there are stark differences across socio-economic groups that are likely to reflect the availability of financial resources in the home. Compared to high SES students, low SES students are more likely to think they will struggle more financially if they go to university (p-value<0.001). They also perceive the difference in terms of having enough money to do what they enjoy as higher (p-value<0.001). The difference in the likelihood of receiving financial support from their parents is significantly lower (p-value=0.020).

As described in Section 2.2, we summarize the 13 variables into six standardized variables for the analysis. Table 3 and Appendix Figure A.7 show mean belief factor scores by SES for each of the six categories. Compared to low SES students, high SES students are more likely to expect to enjoy their social life (0.19 standard deviations) and be interested in the subject/material (0.10 sd). There is a negative but insignificant gap in terms of feeling stressed (-0.06 sd). High SES students are more likely to think they will have parental support in their choice (0.27 sd) and find their life partner

(0.08 sd), and less likely to think they will struggle financially (-0.24 sd).

Table 3: Differences in belief factor scores

	All	SES		
		Low SES	High SES	Difference (p-value)
Social life	0.000 (1.000)	-0.083 (1.027)	0.107 (0.974)	0.191*** (0.000)
Subject interest	0.000 (1.000)	-0.035 (1.033)	0.062 (0.981)	0.097** (0.018)
Stress	0.000 (1.000)	0.044 (1.026)	-0.017 (0.981)	-0.061 (0.135)
Parental support	0.000 (1.000)	-0.126 (0.983)	0.142 (1.001)	0.268*** (0.000)
Life partner	0.000 (1.000)	-0.031 (0.962)	0.044 (1.033)	0.075* (0.068)
Financial struggle	0.000 (1.000)	0.143 (1.019)	-0.101 (0.982)	-0.244*** (0.000)

*Notes.* Column 1 presents the mean across all respondents while columns 2 and 3 show the means for low SES and high SES students, respectively. Column 4 displays the difference in means along with the corresponding p-value testing differences in means. High SES is defined as having at least one parent who has a university degree. The six variables correspond to the extracted factors from the different survey items as described in Section 2.2. \* p<0.10, \*\* p<0.05, \*\*\* p<0.01.

We also find significant differences across socio-economic groups in terms of the perceived later-life returns to university education. Figure 3 and Appendix Table A.6 show differences in perceived returns in terms of (log) earnings, the probability of being employed, and the probability of enjoying the job one will be doing. Both groups perceive the returns to obtaining a university degree and the returns to obtaining high grades to be positive for all later-life outcomes that we measure. Turning to socio-economic differences, most notably, students from low SES backgrounds believe that the earnings boost they will obtain from going to university is significantly lower.<sup>16</sup> They also believe that the additional value of obtaining high grades at university is not as high. While the differences for job-enjoyment premia exhibited in Figure 3 are insignificant, we note that the SES gap in the perceived increase in job enjoyment from

<sup>16</sup>As can be seen in Appendix Figure A.8, which presents earnings and the probability of employment by parental education in levels rather than differences, high SES students perceive earnings to be significantly higher in all three scenarios, with the gap increasing with the level of education. Differences in returns to education in terms of the likelihood of having a job, on the other hand, are driven by high SES students perceiving the likelihood of having a job to be lower if they do not go to university.

getting good grades compared to not going to university is significant. Taken together, students from low SES backgrounds do not only think that they will reach less favorable terminal nodes (Table 2) but they also think that the returns to reaching those nodes is significantly lower.

Figure 3: Perceived later-life returns by SES



Notes: This figure shows the average perceived later-life returns separately by SES. ‘Low grades - No uni’ refers to the perceived difference between the scenario in which the student graduates with low grades and the scenario in which the student obtains no university degree. ‘High grades - Low grades’ refers to the perceived difference between the scenario in which the student graduates with high grades and the scenario in which the student graduates with low grades. Panel A shows differences in perceived earnings, panel B differences in log perceived earnings, panel C differences in perceived employment probabilities and panel D differences in perceived probability of job enjoyment. Stars indicate differences by SES: \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

Before we proceed to the estimation of the choice model, we would like to note that average beliefs about potential earnings in our sample are remarkably close to the earnings statistics we obtain from the 2015 Labour Force Survey (LFS). On average, students in our sample expect to earn £34,194 if they go to university (conditional on being employed), while they expect to earn £23,912 if they start working instead.<sup>17</sup>

<sup>17</sup>The reported averages are computed as weighted averages of expected earnings, where the weights

Using information on individuals in the LFS aged 25-34 who are employed full-time, we document that the actual realized earnings are £33,642 for individuals who went to university while they are £24,752 for individuals who did not go to university. Whether or not realized outcomes differ across socio-economic groups is a question we discuss in Section 8.

## 5 The Dynamic Choice Model

The aim of this study is to understand why students decide to attend university and why we observe such a large SES gap in enrollment. In order to model this important life decision, we treat the decision to enroll as the first stage of a dynamic choice problem in which students are forward-looking and take future decisions and their associated returns into account when making contemporaneous choices. Students differ across several key dimensions. First, they differ in terms of gender and socio-economic background, both assumed to be constant. Second, students hold heterogeneous beliefs about the pecuniary and non-pecuniary returns to university attendance. Some of these returns relate to the university experience itself, while others relate to returns accruing later in life on the labor market. Finally, students hold different beliefs about their own ability and probability of obtaining good grades in case of university completion. While all of the previously mentioned characteristics and beliefs differ across individuals, for notational convenience, we suppress indexing them by  $i$  in what follows.

Students from different socio-economic backgrounds might not only hold different beliefs, they may also have different preferences. On the one hand, preferences might differ in terms of the level of utility students derive from choosing a specific alternative. Our model allows for such level differences across socio-economic groups. On the other hand, differences in preferences might go deeper than that. In particular, there might be differences in terms of the weight low and high SES students place on different aspects of each decision. For example, students from low SES background may place more or 

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are the perceived probabilities of reaching a given terminal node.

less weight on non-pecuniary factors when making their decision of whether to enroll in university. We use a flexible model specification in which we allow all preference parameters to vary by SES by estimating the model separately for high and low SES students. For notational convenience we also suppress this group-level heterogeneity in the exposition of the model.

Students face either two or three sequential decisions, depending on whether they initially decide to enroll in university. More specifically, students face the following choices: first, students face the choice  $s$  between not going to university but starting to work instead ( $s_0$ ), enrolling in university while working alongside their studies ( $s_1$ ), or enrolling in university but not working alongside their studies ( $s_2$ ); second, conditional on enrollment students face the decision  $c$  whether to complete university ( $c_1$ ) or drop out ( $c_0$ ); and third, at age 30, all individuals face the choice  $l$  whether to work ( $l_1$ ) or not ( $l_0$ ). Strictly speaking, individuals contemplate three different work decisions  $l^d$  across the final decision nodes of not obtaining a university degree ( $d = 0$ ), or obtaining a university degree with lower ( $d = 1$ ) or higher grades ( $d = 2$ ). For notational convenience, we will occasionally summarize the three work choices by  $l$ . Let the set of all choices be denoted by  $j \in \{s, c, l\}$ .

At each of these decision nodes, individuals draw independently distributed random utility shocks  $\varepsilon_j$  for each option, each of which is distributed according to an extreme value distribution. These unobserved state variables can be interpreted as transitory and idiosyncratic shocks to the utility from the respective decisions. Since only differences in utilities are of importance in this type of decision problem, we define  $\varepsilon_{j_0} \equiv 0$  so that  $\varepsilon_j$  is relative to option zero.

In addition to the shocks  $\varepsilon_j$ , which are not known to the individual until the respective decision node is reached, two students with identical characteristics and beliefs may still make different choices due to additive utility components which are known to the individuals *ex ante* but are not observable to the researcher. In order to capture this possibility, we add unobserved heterogeneity in the form of additive components  $\xi_j$  for  $j \in \{s_1, s_2, c, l\}$  which are assumed to be normally distributed with mean zero.



For the sake of dimensionality reduction, individuals only possess one single unobserved component  $\xi_l$  across the three possible work decision nodes.<sup>18</sup> The variance of the unobserved heterogeneity is normalized to one for the employment decision ( $l$ ), while the variances (and covariances) of the remaining three components are estimated.

In what follows each decision is assigned a time subscript  $t \in \{0, 1, 2\}$ . Let us assume that expected utility of each subsequent decision is discounted at rate  $\beta$ . The corresponding value functions are denoted by  $V_t^j$  of decision  $j$ . Moreover, values of choosing  $j = 1$  are indicated by  $V_t^{j1}$ , while values of choosing  $j = 0$  are represented by  $V_t^{j0}$ . The direct utility for each period and choice will be denoted analogously by  $U_j$ , with the subscript indicating decision  $j$ . Since only differences in utility matter, the direct utilities of not enrolling, not completing, and not working at age 30 are assumed to be zero, i.e.  $U_{s_0} = U_{c_0} = U_{l_0^d} = 0$ .  $X_j$  is the vector of individual characteristics pertinent to the direct utility associated with decision  $j$ . Given that with probability  $p_j$  an individual chooses  $j$ , future values are convex combinations of  $V_t^{j1}$  and  $V_t^{j0}$ ; i.e.  $V_t^j = p_j V_t^{j1} + (1 - p_j) V_t^{j0}$ .

## 5.1 The Enrollment Decision

Attending university is associated with different immediate non-pecuniary benefits and costs that are perceived differently by each individual. If an individual decides to attend university during the 3-4 years after finishing high school, she derives utility from  $N$  different non-pecuniary benefits/costs  $q_n$  associated with university life, each of which is weighted by a corresponding parameter  $\theta_n$ . The immediate non-pecuniary benefits/costs of university attendance relate to students' (i) social life, (ii) interest in the subject, (iii) stress, (iv) parental support in the choice, and (v) finding a life partner (see Section 2.2 for details). For each of these non-pecuniary benefits/costs,  $q_n$  is defined as the difference in the benefits/costs if the student attends university vs. if

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<sup>18</sup>Conceptually, one can also imagine that an unobserved (dis)taste of work is independent of which type of job is faced.

the student starts working instead.<sup>19</sup>

At the first node, the student decides between not attending university, attending university and working alongside her studies, or attending university but not working alongside her studies. This decision is influenced by her perceived financial situation as well as other factors. We allow some of the non-pecuniary benefits/costs of university education to vary with the decision whether to work alongside her studies or not. In particular, we allow students' social life, interest in subject, and stress to have an endogenous component affected by the choice of whether or not to work. Conceptually, we do not think that parental support in the students' choice to go to university or the probability of finding a life partner should be seen as a function of the decision to work alongside their studies, which is why we do not let these benefits/costs to be affected by the decision. We denote the potential non-pecuniary benefits/costs derived from university attendance if working alongside as  $q_n^{s1}$  and if not working alongside as  $q_n^{s2}$ . While perceived non-pecuniary benefits/costs are allowed to differ across all individuals, the difference between the potential non-pecuniary benefits/costs when not working alongside studies versus working alongside studies ( $\Delta q_n \equiv q_n^{s1} - q_n^{s2}$ ) is assumed to be the same for all individuals, but is allowed to differ from one aspect  $n$  to another. While working alongside studies might have a large influence on whether a student feels stressed or not, it might only have a minor impact on whether the student enjoys the subject she studies. Motivated by cross-sectional evidence in our data, we assume no effect of working alongside studies on perceived completion probabilities. We also assume that working alongside studies does not affect expected wages, for example, through the accumulation of occupation specific human capital. We summarize the potential non-pecuniary benefits/costs of university attendance when working alongside university by vector  $Q^{s1}$  and when not working alongside university by  $Q^{s2}$ , while the vector of relative weights on each item  $n$  is denoted as  $\Theta$ .

Including the decision to work alongside studies in the choice model is motivated

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<sup>19</sup>The variables  $q_n$  are standardized to have a mean of zero and a standard deviation of one, as explained in Section 2.2.

by the following three patterns in the data: (i) according to a nationally representative survey of university students in the UK, a large majority of university students (77%) works for a considerable number of hours alongside their studies (Endsleigh Student Survey 2015);<sup>20</sup> (ii) our data reveals that there is a sizeable socio-economic gap in students' intentions to work alongside their studies (see Section 2); (iii) students' intentions to work are significantly related to how they perceive their financial situation. At the same time, students who intend to work while studying believe they are less likely to enjoy their social lives, less likely to enjoy their studies, and more likely to feel stressed. We provide further details in Section 5.5.

A student makes her decision given the direct utility and future values of either choice. At  $t = 0$  the student will choose between the three options of not going to university, going to university and working alongside her studies, or going to university and not working alongside her studies according to the following:

$$V_0 = \max_{s_j} [s_1 U_{s_1}(X_s) + s_2 U_{s_2}(X_s) + E[V_1^s]] \quad (1)$$

where in case of enrollment the student enjoys direct utility for each  $j \in 1, 2$

$$U_{s_j}(X_s) = \kappa_{s_j} + \Theta Q^{s_j} + \mu_{s_j} * male + \psi_{s_j} * fin\_struggle + \varepsilon_{s_j} + \xi_{s_j}. \quad (2)$$

We assume that the direct utility derived from attendance while working alongside her studies,  $s_1$ , and without working alongside her studies,  $s_2$ , have intercepts  $\kappa_{s_1}$  and  $\kappa_{s_2}$ , and change for males by  $\mu_{s_1}$  and  $\mu_{s_2}$ , respectively. Students worrying about struggling financially value these concerns relative to not attending university with parameters  $\psi_{s_1}$  and  $\psi_{s_2}$ . The random utility shocks are denoted by  $\varepsilon_{s_1}$  and  $\varepsilon_{s_2}$ , while  $\xi_{s_1}$  and  $\xi_{s_2}$  capture unobserved heterogeneity across individuals.

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<sup>20</sup>Students earn about £412 per month, which, assuming they earn the minimum wage, would imply that students work about 15 hours/week .

## 5.2 The Completion Decision

If a student does not enroll in university, then no completion decision is to be made. The value at period  $t = 1$ , if the student has not enrolled in university, is the future discounted value at age 30:

$$V_1^{s0} = \beta E[V_2^{c0}], \quad (3)$$

i.e. the utility in the intermediate period is normalized to zero and the next decision the student will face is whether to work at age 30 or not.

If enrolled in university, then the student needs to decide  $c$  in period  $t = 1$ , i.e. whether to complete university or drop out:

$$V_1^{s1} = V_1^{s2} = \max_c [cU_c(X_c) + \beta E[V_2^c]] \quad (4)$$

where the direct utility associated with completion  $c$  is given by

$$U_c(X_c) = \kappa_c + \mu_c * male + \alpha_c * ability + \varepsilon_c + \xi_c. \quad (5)$$

The direct utility from completion has an intercept  $\kappa_c$ , changes for males by  $\mu_c$ , and might be increase by  $\alpha_c$  for a student with ability due to lower effort costs.  $\varepsilon_c$  is the random utility shock, while  $\xi_c$  captures unobserved heterogeneity across individuals.

In case of university completion, a student receives the news  $\gamma$  whether she graduated with higher ( $\gamma_G$ ) grades, which occurs with her perceived probability  $p$ , or lower grades ( $\gamma_B$ ) with probability  $(1 - p)$ . This will affect wages and the likelihood of her enjoying her job (if she works). Given the student does not know her grades in the case of completion, she uses her perceived probability of graduating with high or low grades to form expectations according to:

$$E[V_2^{c1}] = pE[V_2^2] + (1 - p)E[V_2^1] \quad (6)$$

while in case of not attending university or not graduating

$$V_2^{co} = E[V_{l^0}]. \quad (7)$$

### 5.3 The Work Decision

If an individual works at age 30, she derives utility  $\lambda$  from enjoying the job and values earnings  $w_d$  by the function  $v(w_d) = \log(w_d)$  with associated weight  $\tau$ . Individuals hold beliefs concerning the likelihood of enjoying their job  $\pi_d$  and expected wages  $w_d$  across the three final decision nodes of not obtaining a university degree ( $d = 0$ ), or obtaining a university degree with lower ( $d = 1$ ) or higher grades ( $d = 2$ ).

The direct utility in period  $t = 2$  depends on whether the individual decides to work ( $l = 1$ ) or not ( $l = 0$ ) and can be summarized by:

$$V_{l^d} = \max_{l^d} [l^d(\tau v(w_d) + \lambda \pi_d) + U_{l^d}(X_l)] \quad (8)$$

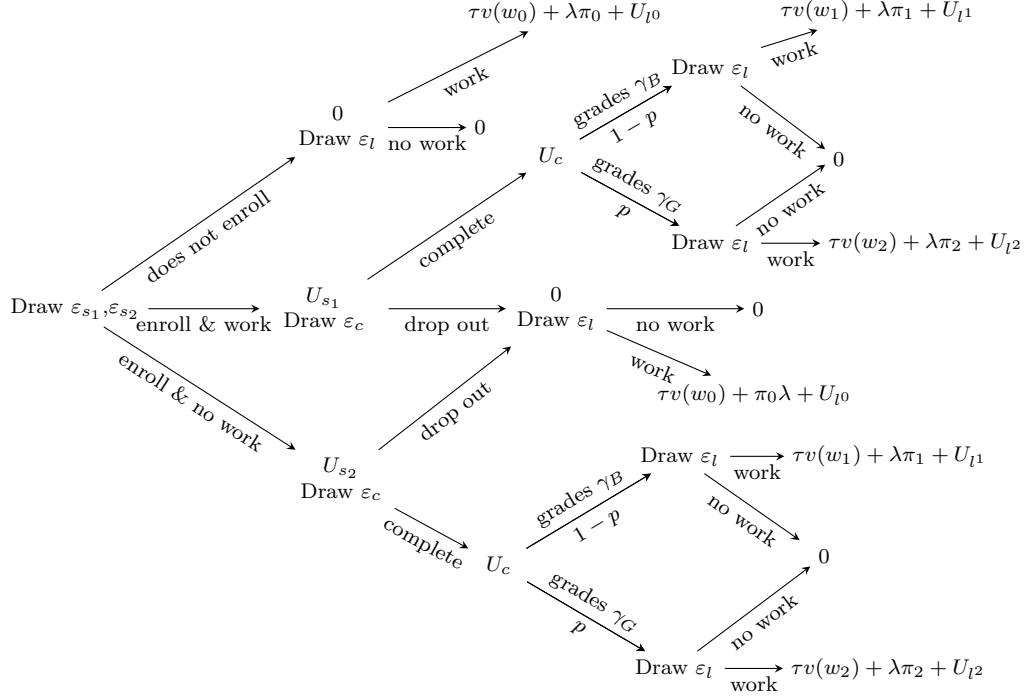
where  $v(w) = \log(w)$ ,  $\tau$  is the relative weight attached to perceived monetary returns, and the direct utility is

$$U_{l^d}(X_l) = \kappa_{l^d} + \alpha_l * \nu + \varepsilon_{l^d} + \xi_l. \quad (9)$$

$\kappa_{l^d}$  are the intercepts and  $\nu$  is the individual propensity to work, which is estimated outside the model and discussed in the following section.  $\varepsilon_{l^d}$  is the random utility shock, while  $\xi_l$  captures unobserved heterogeneity across individuals. The utility associated with not working is assumed to be zero.

Figure 4 displays all decision nodes, shocks as well as the associated utility an individual derives.

Figure 4: Decision tree



## 5.4 Estimation and Identification

As outlined in Figure 4 and Appendix C, solving the full dynamic programming problem involves five decisions for each individual, which are associated with the following probabilities of choosing a specific alternative:

1. Probabilities of working  $p_{ld}$  under three scenarios:
  - (a) No university degree ( $d = 0$ ).
  - (b) University degree with lower grades ( $d = 1$ ).
  - (c) University degree with high grades ( $d = 2$ ).
2. University completion probability  $p_c$ .
3. Enrollment probability working alongside studies  $p_{s1}$  or not working alongside studies  $p_{s2}$ .

We write the probability of enrolling in university as  $p_s \equiv p_{s_1} + p_{s_2}$ , i.e. as the sum of the probability of going to university and working and the probability of going to university but not working alongside studies. The probability of working alongside studies conditional on going to university is denoted as  $p_w$ . The model contains 20 free parameters which determine these decision probabilities and 9 additional free parameters for the variance-covariance matrix of the unobserved heterogeneity. In the following, we outline which parameters are determined outside the model and how the remaining parameters are estimated through simulated method of moments.

## 5.5 Parameterization

For all decisions, we include an intercept  $\kappa_j$  to allow for baseline levels of utility to differ across alternatives. For the initial choice of whether to attend university and work alongside as well as for the choice of university completion, we additionally include parameters  $\mu_j$  to allow for the level of utility derived from enrollment and completion to differ across gender. We do not include the gender dummy for the decisions to work at the final node. For the utility derived from work, we can make use of the fact that we have three different scenarios  $d$  (no degree, university with lower grades, university with good grades) for which we know each individual’s perceived probability of working. Therefore, outside the model, we run a regression pooling all students and adding individual fixed effects in order to gauge each student’s individual affinity to work given a wage, job enjoyment level, and scenario fixed effect  $\chi_d$ . More specifically, the regression takes the form:

$$p_{td} = \nu + \beta_1 * w_d + \beta_2 * \lambda_d + \chi_d + \varepsilon. \quad (10)$$

We can then use the estimated individual fixed effects  $\nu$  as a proxy for each student’s willingness to work.

Working alongside university might affect students’ non-pecuniary benefits/costs from attending university. Given that we did not elicit students’ perceived bene-

fits/costs for both scenarios, i.e. working alongside university versus not working alongside university, we derive the impact from the cross-sectional variation by regressing individual perceived benefits/costs on the probability of working alongside university conditional on enrollment,  $p_w$ , as well as controls. The controls include a constant, gender and SES dummies, the perceived probability of obtaining the necessary grades, and school fixed effects. Throughout our analysis, we use the stated probability of attaining the necessary grades to attend university as a proxy for perceived ability.<sup>21</sup> The regression takes the following form:

$$q_n = \beta_0 + \beta_1 * p_w + \beta_2 * male + \beta_3 * SES + \beta_4 * ability + \chi + \varepsilon \quad (11)$$

where  $\chi$  are school fixed effects. In Table 4 we summarize the impact of working while studying on the perceived experiences. The impact can be interpreted in terms of standard deviations. More specifically, working alongside university is estimated to reduce the perceived non-pecuniary benefit of university attendance in terms of social life by 0.36 standard deviations, reduce the perceived non-pecuniary benefit in terms of enjoyment of the subject by 0.28, and increase the perceived non-pecuniary costs in terms of stress by 0.21.

From this exercise we can construct counterfactual non-pecuniary benefits/costs of attending university for each individual depending on whether or not the individual works alongside her studies:

$$\begin{aligned} q_n^{s_2} &= q_n - \beta_1 * p_w && \text{if } s_2 = 1 \\ q_n^{s_1} &= q_n + \beta_1 * (1 - p_w) && \text{if } s_1 = 1 \end{aligned} \quad (12)$$

where  $p_w$  is the stated individual probability of working alongside university and  $q_n$  is the extracted factor for item  $n$  concerning the experience at university relative to not

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<sup>21</sup>We find that the stated probability of attaining necessary grades correlates highly with Math and English grades. However, we only observe grades for a relatively small sample, which is why we use the probability of attaining necessary grades as a measure for ability instead. We would like to note that there is no significant SES gap in the probability of attaining necessary grades once we control for test scores.



Table 4: Effect of working part-time on immediate non-pecuniary factors

Dependent variable: Costs/benefits of university attendance			
	(Social life)	(Enjoy subject)	(Stress)
Work alongside studies (0-1)	-0.36*** (0.08)	-0.28*** (0.07)	0.21*** (0.08)
Controls	Yes	Yes	Yes
$R^2$	0.092	0.126	0.053
Observations	2356	2356	2356

Notes: \* $p < 0.10$ , \*\* $p < 0.05$ , and \*\*\* $p < 0.01$ . The estimation technique is OLS. Standard errors are in parenthesis. The dependent variable is indicated in the top of the column. The effect size is in terms of standard deviations of the respective extracted factor. The controls include a constant, gender and SES dummies, the perceived probability of obtaining the necessary grades, and school fixed effects. Work alongside studies is the perceived probability of working alongside university, measured on a 0-1 scale.

attending.

Assuming that the yearly discount rate is about 0.96 and age 30 follows about 5-6 years after graduation, we use a discount rate  $\beta$  of 0.80 for expected utility at age 30.

We summarize the parameters fixed outside the model in Table 5.

Table 5: Parameters set outside the model

Parameter	Description	Values
$\beta$	Discount rate	0.8
$\Delta q_1$	Difference in social life at university when working alongside	-0.36
$\Delta q_2$	Difference in interest in subject at university when working alongside	-0.28
$\Delta q_3$	Difference in stress at university when working alongside	0.21

For each socio-economic group, the model contains 29 free parameters; 20 preference parameters summarized in Table 6 and nine parameters for the variance-covariance matrix of the unobserved heterogeneity. To pin down the free parameters, we target 70 data moments for each socio-economic group related to the five decisions, which are computed using the microdata. In order to identify parameters related to each decision in a static sense, we target the mean expected choice, and correlations between choices and expected immediate returns. To add a dynamic component, following Eisenhauer, Heckman and Mosso (2015), we target regression coefficients of future returns on con-

Table 6: Parameters estimated inside the model

	N	Description
<i>Enrollment</i>		
$\Theta$	5	Utility gain/loss from experiences at university
$\psi_1$	1	Utility cost of financial struggles when working alongside studies
$\psi_2$	1	Utility cost of financial struggles when not working alongside studies
$\mu_{s1}$	1	Male preference for enrollment and working alongside studies
$\mu_{s2}$	1	Male preference for enrollment w/o working alongside studies
$\kappa_{s1}$	1	Preference for enrollment and working alongside studies
$\kappa_{s2}$	1	Preference for enrollment w/o working alongside studies
<i>Completion</i>		
$\alpha_c$	1	Utility from ability in case of completion
$\mu_c$	1	Male preference for completion
$\kappa_c$	1	Preference for completion
<i>Work</i>		
$\tau$	1	Relative weight on monetary returns
$\lambda$	1	Utility gain from job one enjoys
$\alpha_l$	1	Utility from individual work propensity in case of working
$\kappa_{l0}$	1	Preference for work (no degree)
$\kappa_{l1}$	1	Preference for work (low grades)
$\kappa_{l2}$	1	Preference for work (high grades)
<i>Other</i>		
$\xi$	9	Unobserved heterogeneity (variance-covariance matrix)
Total	29	

temporaneous decisions. For instance, beliefs about the university premium might play a role when deciding whether or not to go to university. For this purpose, we regress expected choices on items influencing the direct utility as well as future returns. The full list of moments we target is presented in Table 7.

Note that for identification of the preference weights on perceived returns, we have both the correlation coefficients between perceived returns and decisions as well as the OLS coefficients. Given the potential importance of gender, gender dummies are identified by OLS coefficients and mean decisions by males. Hence we have at least two, albeit not fully independent, targeted moments to discipline each of these parameters. Utility intercepts are identified by mean decisions.

Table 7: Targeted moments

Moments	N
<i>Means</i>	
Mean choice of all six decisions	6
Mean choice of enrolling and working alongside studies	1
Mean choice by males of all six decisions	6
<i>Correlations</i>	
Correlation between wages and work choice for all three work nodes	3
Correlation between job enjoyment and work choice for all three work nodes	3
Correlation between work propensity and work choice for all three work nodes	3
Correlation between ability and completion	1
Correlation between probability of good grades and completion	1
Correlation between non-pecuniary returns at uni and enrollment	5
Correlation between financial struggles and enrollment	1
Correlation between financial struggles and working alongside studies	1
<i>Auxiliary regressions</i>	
OLS coefficients of enrollment without working alongside regressed on non-pecuniary returns to uni, ability, probability of good grades, gender, financial struggles, expected job enjoyment, and expected wages	11
OLS coefficients of enrollment and working alongside regressed on non-pecuniary returns to uni, ability, probability of good grades, gender, financial struggles, expected job enjoyment, and expected wages	11
OLS coefficients of completion regressed on ability, probability of good grades, gender, expected job enjoyment, and expected wages	5
OLS coefficients of work decisions regressed on gender, wages, job enjoyment, and propensity to work	3x4
Total	70

In order to estimate the parameters, we use simulated method of moments in which we minimize the weighted distance between data moments and moments generated by the model. One advantage of using SMM in our context is that we do not require ad-hoc transformations of zeros and ones. Written formally, let  $\mathbf{M}$  represent the vector of these 70 moments. A vector of the analogous 70 moments can be obtained from the estimated model. The moments for the model are a function of the parameters to be estimated. Let  $\mathcal{M}(\mathbf{P})$  represent this vector of moments, where  $\mathbf{P}$  denotes the vector of 29 parameters to be estimated. Define the vector of deviations between the data and the model by  $\mathbf{G}(\mathbf{P}) \equiv \mathbf{M} - \mathcal{M}(\mathbf{P})$ . Minimum distance estimation picks the parameter vector,  $\mathbf{P}$ , to minimize a weighted sum of the squared deviations between the data and

the model, i.e.,

$$\hat{\mathbf{P}} = \arg \min \mathbf{G}(\mathbf{P})' \mathbf{W} \mathbf{G}(\mathbf{P}).$$

The estimated parameter vector  $\hat{\mathbf{P}}$  is consistent for any semi-definite matrix  $\mathbf{W}$ . As a weighting matrix, we choose the diagonal matrix containing the inverse variances of the respective moments based on 10,000 bootstrapped draws. The variance-covariance matrix for  $\hat{\mathbf{P}}$  is consistently estimated as

$$\text{var}(\hat{\mathbf{P}}) = \left[ D(\hat{\mathbf{P}})' \mathbf{W} D(\hat{\mathbf{P}}) \right]^{-1} D(\hat{\mathbf{P}})' \mathbf{W} \hat{\mathbf{Q}} \mathbf{W} D(\hat{\mathbf{P}}) \left[ D(\hat{\mathbf{P}})' \mathbf{W} D(\hat{\mathbf{P}}) \right]^{-1}, \quad (13)$$

where  $D(\hat{\mathbf{P}})$  is a matrix of partial derivatives of the moments included in  $\mathcal{M}(\mathbf{P})$  with respect to the parameters included in  $\mathbf{P}$ , and  $\hat{\mathbf{Q}}$  is an estimate of the variance-covariance matrix of the moments in the data.<sup>22</sup> Given the extreme value distribution of the random utility shocks, the decisions can be written in logit-form. The unobserved heterogeneity is integrated out using 100 draws from a Halton sequence.

## 6 Results

### 6.1 Choice Model Estimates

We separately estimate the model for low SES students ( $N = 1055$ ) and high SES students ( $N = 1301$ ). Table 8 presents the results for low (left) and high (right) SES students. All variables have been standardized to have a mean of zero and a standard deviation of one, so the coefficients can be interpreted relative to each other. We see that students place significant value on non-pecuniary returns of university attendance when deciding upon enrolling, amongst which subject interest and parental support play the largest role. For both high and low SES students, parental support in their choice plays the most important role. In contrast, concerns about social life do not significantly enter students' decisions and finding one's life partner during the course

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<sup>22</sup>When we estimate  $\hat{\mathbf{Q}}$ , we include both variance and covariance terms. Our procedure follows, among others, Llull (2018), equation A42.

of one's studies, if anything, is valued negatively by high SES students. Worries about financial struggles make students less likely to want to enroll in university without working alongside their studies. Boys are significantly less likely to enroll in university and work alongside their studies, whereas for enrollment without part-time work the coefficient is negative but insignificant. Despite having detailed information on beliefs about returns, we find that factors not captured by beliefs but correlated with SES still play a significant role in the enrollment decision, which is indicated by the larger intercepts for high SES students.

Table 8: Estimated parameters

Description	Parameter	Low SES		High SES	
		Value	95% CI	Value	95% CI
<i>Enrollment</i>					
Social life	$\theta_1$	-0.024	[-0.065, 0.017]	-0.021	[-0.054, 0.011]
Subject interest	$\theta_2$	0.056	[0.011, 0.101]	0.038	[0.004, 0.071]
Stress	$\theta_3$	-0.049	[-0.104, 0.006]	-0.06	[-0.096, -0.024]
Parental support in choice	$\theta_4$	0.093	[0.051, 0.134]	0.06	[0.029, 0.09]
Finding partner	$\theta_5$	0.011	[-0.031, 0.053]	-0.036	[-0.069, -0.004]
Financial struggle work	$\psi_1$	-0.028	[-0.074, 0.019]	0.006	[-0.029, 0.042]
Fin. struggle no work	$\psi_2$	-0.083	[-0.133, -0.032]	-0.037	[-0.075, 0.001]
Male enroll and work	$\mu_{s_1}$	-0.465	[-0.67, -0.26]	-0.376	[-0.574, -0.177]
Male enrollment	$\mu_{s_2}$	-0.085	[-0.325, 0.156]	-0.117	[-0.335, 0.101]
Intercept enroll and work	$\kappa_{s_1}$	-0.034	[-0.174, 0.105]	0.134	[0.005, 0.264]
Intercept enrollment	$\kappa_{s_2}$	-1.301	[-1.451, -1.151]	-0.538	[-0.676, -0.401]
<i>Completion</i>					
Ability	$\alpha_c$	0.155	[0.111, 0.199]	0.145	[0.109, 0.18]
Male completion	$\mu_c$	-0.125	[-0.26, 0.01]	-0.05	[-0.178, 0.078]
Intercept completion	$\kappa_c$	1.005	[0.896, 1.114]	1.045	[0.951, 1.139]
<i>Work</i>					
Wage	$\tau$	0.457	[0.432, 0.482]	0.472	[0.456, 0.489]
Job enjoyment	$\lambda$	0.637	[0.602, 0.673]	0.605	[0.573, 0.638]
Propensity to work	$\alpha_l$	1.106	[1.075, 1.138]	1.107	[1.077, 1.137]
Intercept working (no degree)	$\kappa_{l0}$	1.854	[1.747, 1.96]	1.73	[1.628, 1.833]
Intercept working (low grades)	$\kappa_{l1}$	1.77	[1.675, 1.866]	1.823	[1.729, 1.917]
Intercept working (high grades)	$\kappa_{l2}$	2.298	[2.192, 2.403]	2.267	[2.175, 2.359]

Notes: The estimation technique is SMM. Confidence intervals are computed using the Delta method. The model is estimated separately for low and high SES students.

For the completion decision, we find that perceived ability plays a significant and similar role for both low and high SES students. Completion, in general, seems to be valued similarly across SES as indicated by the comparable intercepts. Boys seem to place a lower value on completion than girls. This gender gap is statistically significant for low SES students but not for high SES students.

Concerning the decision to work, job enjoyment plays a greater role than wages for both high and low SES students. While high SES students tend to place slightly more weight on wages than low SES students, the opposite holds true for job enjoyment. Concerning the general valuation of work under the three different scenarios, we find that low SES students value work with no degree more than high SES students do, while we do not see substantial differences for work with low or high grades.

Summarizing the results from the estimation, we see that low SES students tend to place more weight on non-pecuniary factors such as subject interest, parental support, and job enjoyment, while high SES students seem to care more about expected wages and enrollment in general.

In Table 9 we present the fit for the main decisions in the model. Data moments are on the left and model moments on the right followed by the 95% confidence intervals obtained by simulating 1,000 bootstrapped draws from the respective samples with parameters drawn from the variance-covariance matrix computed using the Delta method. In the top panel of Table 9 we present the fit for low SES (left) and high SES students (right) and add a range of unconditional moments which were not targeted directly. The following panels contain the sample of low SES (middle) and high SES students (bottom) broken down by girls (left) and boys (right). For all of the conditional decisions, which are the ones we target, the data moment lies within the relatively tight 95% confidence interval around the model moment.

Table 9: Model fit by SES

Decision	Data	Model	95% CI	Data	Model	95% CI
	Low SES			High SES		
Enroll	0.742	0.741	[0.722, 0.76]	0.831	0.83	[0.815, 0.843]
Work alongside studies	0.705	0.707	[0.693, 0.72]	0.633	0.633	[0.618, 0.647]
Complete	0.817	0.812	[0.799, 0.824]	0.85	0.849	[0.839, 0.858]
Complete (unconditional)	0.634	0.614	[0.593, 0.634]	0.726	0.712	[0.695, 0.727]
Complete (low grades)	0.191	0.2	[0.19, 0.209]	0.188	0.196	[0.188, 0.203]
Complete (high grades)	0.442	0.414	[0.399, 0.432]	0.539	0.516	[0.501, 0.531]
Work no uni	0.715	0.715	[0.698, 0.734]	0.677	0.68	[0.662, 0.697]
Work uni low grades	0.781	0.778	[0.762, 0.793]	0.774	0.777	[0.763, 0.791]
Work uni high grades	0.861	0.86	[0.848, 0.873]	0.863	0.861	[0.85, 0.872]
Work no uni (uncond.)	0.262	0.271	[0.255, 0.288]	0.181	0.192	[0.18, 0.205]
Work uni low grades (uncond.)	0.148	0.157	[0.148, 0.165]	0.144	0.152	[0.145, 0.159]
Work uni high grades (uncond.)	0.392	0.37	[0.354, 0.388]	0.479	0.457	[0.442, 0.472]
$\chi^2$	5.18 (0.952)			4.87 (0.962)		
<i>Low SES</i>						
	Girls			Boys		
Enroll	0.771	0.769	[0.745, 0.79]	0.691	0.694	[0.664, 0.725]
Work alongside studies	0.724	0.73	[0.712, 0.748]	0.672	0.666	[0.643, 0.691]
Complete	0.826	0.818	[0.803, 0.832]	0.802	0.801	[0.781, 0.82]
Work no uni	0.698	0.697	[0.676, 0.72]	0.744	0.746	[0.719, 0.77]
Work uni low grades	0.761	0.763	[0.745, 0.781]	0.816	0.803	[0.782, 0.824]
Work uni high grades	0.849	0.852	[0.838, 0.867]	0.88	0.874	[0.857, 0.892]
$\chi^2$	0.09 (>0.999)			0.12 (>0.999)		
<i>High SES</i>						
	Girls			Boys		
Enroll	0.837	0.835	[0.817, 0.851]	0.821	0.821	[0.797, 0.842]
Work alongside studies	0.642	0.653	[0.635, 0.67]	0.617	0.599	[0.574, 0.623]
Complete	0.844	0.845	[0.832, 0.856]	0.859	0.856	[0.842, 0.869]
Work uni low grades	0.756	0.759	[0.743, 0.776]	0.805	0.81	[0.791, 0.827]
Work uni high grades	0.85	0.849	[0.836, 0.86]	0.886	0.883	[0.868, 0.895]
$\chi^2$	0.2 (>0.999)			0.28 (>0.999)		

Notes: Unless specified otherwise, decisions are conditional on arriving at the respective node. Unconditional moments are not targeted directly. Confidence intervals are computed based on 1,000 bootstrapped samples and random parameter draws from the estimated parameter variance-covariance matrix.  $\chi^2$ -statistics of goodness of fit are presented at the bottom of each panel with p-values in brackets.

As proposed in Heckman (1984) and Heckman and Walker (1990), we evaluate the goodness of fit using  $\chi^2$ -tests. In the top panel of Table 9, which includes six targeted and six not directly targeted moments, the model predictions are such that we cannot reject the null that the data moments and the simulated moments are the same at any

conventional level. The  $\chi^2$ -statistics are 5.18 and 4.87 for low and high SES students, respectively, which are substantially lower than the critical value for a test at the 10% significance level. If we only investigate the model fit for the six targeted moments, the  $\chi^2$ -statistics drop to 0.05 and 0.06, respectively, indicating a very high model fit. Similarly, when breaking down low and high SES students by gender, we find very low values for the  $\chi^2$ -statistics that do not allow us to reject the null that the data moments and the simulated moments are the same. Finally, we investigate the goodness of fit across all moments presented in Table 9. We obtain a  $\chi^2$ -statistic of 10.74 which again does not allow us to reject the null at any conventional level.<sup>23</sup>

## 6.2 Model Validation

Despite the model clearly providing a close match to decisions taken by students, we conduct a range of exercises to increase the confidence in the model’s validity by following different steps proposed in Eisenhauer, Heckman and Mosso (2015). First, we begin the estimation algorithm with random initial values for the parameters in order to safeguard against local optima. We find that this makes no difference for the parameters the algorithm settles on. Second, we simulate students’ decisions based on the estimated parameters and then re-estimate the model on the simulated data. We recover the same parameters as in the benchmark model. Third, in Appendix Figure C.1 we plot the behavior of the criterion function when each parameter is shifted by up to two standard deviations to the left and right from the optimum. We see that the criterion function behaves smoothly for each parameter further increasing our confidence that we have found the global minimum.

In order to obtain a better sense of the model’s sensitivity to each parameter, we simulate the model while shifting parameters by 0.5 standard deviations and look at the impact on the five conditional decisions students make in the model in Appendix Figure C.2. The results of this exercise are discussed in detail in Appendix C.3. In

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<sup>23</sup>We correct the number of degrees of freedom for the fact that when including the shares of girls, the shares of boys are linearly dependent when combined with the aggregate shares from the top panel.



summary, the sensitivity analysis reveals several model characteristics. First, no single parameter dominates decisions, i.e. shifting any single parameter by 0.5 standard deviations does not lead to dramatic shifts in decisions. Second, while shifts are not dramatic, each parameter at least affects the respective decision, thereby strengthening the identification argument. Third, dynamic considerations become very clear as changing returns to decisions later in life, e.g. the work decision, trickles down to changes in decisions earlier in life, e.g. enrollment. For example, an increase in the value placed on job enjoyment  $\lambda$  increases the probability of working with a high degree, because expected job enjoyment tends to be greater for this case, whereas the probability of working with no university degree decreases, as expected job enjoyment tends to be smaller in this situation. This shift in preferences trickles down to earlier decisions as well. Now that high-degree jobs have become more attractive, students become more likely to complete university. This, in turn, also increases the probability of enrolling in university in the first place.

## 7 Simulations

In the following, we examine to what extent beliefs about pecuniary and non-pecuniary returns contribute to the decision to enroll in university, and what share of the SES gap in enrollment can be attributed to differences in beliefs versus preferences.

### 7.1 Why Do Students Go to University?

In order to gauge the importance of differences in beliefs across students in their decisions to go to university, we run counterfactual simulations in which we assign average beliefs to all individuals in the sample. We then examine the model performance for the enrollment decision in terms of the correlation between data and model enrollment probabilities. We present the relative performance to the benchmark, which is scaled to 100, in Table 10.

Table 10: Importance of perceived returns for enrollment decision

	Correlation
Benchmark	100
All returns	55.0
Non-pecuniary returns	69.6
Non-pecuniary returns at uni	86.0
Pecuniary returns	98.2
Labor returns	77.5
Ability	85.3

Notes: Each row stands for an experiment in which we assign mean beliefs to all individuals in the model. The left column specifies which beliefs are altered. The second column represents the correlation between data and model enrollment probabilities. All values are relative to the benchmark which is normalized to 100.

Disregarding variation in students' beliefs about all of the pecuniary and non-pecuniary returns we measure reduces the correlation by 45%. Disregarding variation in beliefs about non-pecuniary returns alone reduces the correlation by 30%, while disregarding variation in beliefs about pecuniary returns reduces the correlation by only 2%. We find that differential beliefs about ability play an equally important role as perceived non-pecuniary returns that relate to students' experiences at university, both accounting for about 15% of the model's capability to explain enrollment decisions. Finally, disregarding differences in perceptions about all labor market returns reduces the correlation by 22%.

## 7.2 SES Gap - Differences in Beliefs

To gain a better understanding of the extent to which differences in beliefs contribute to the socio-economic gap in enrollment, we first simulate counterfactuals in which low SES students draw from the belief distribution of high SES students. More specifically, we compute the mean and variance of high SES students' beliefs concerning a particular perceived return, e.g. perceived stress. Then for each low SES student, we draw a belief from a normal distribution with the corresponding mean and variance and simulate decisions. We repeat this procedure 1,000 times and present the average share of the SES gap which has been closed and the corresponding 95% confidence intervals.

In Table 11 we present the results and see that 25% of the SES gap in enrollment can be explained by differences in perceived returns. Non-pecuniary returns account for 13%, while pecuniary returns for 15%. Amongst non-pecuniary returns at university, parental support (6%) and subject interest (7%) contribute most to the gap. The former suggests that parents influence children's decisions to obtain further education through more than only material support but also through emotional support in their decision. The latter indicates that children from more educated backgrounds have a greater thirst for obtaining knowledge in an academic environment. Taken together, the importance of the differences in beliefs across these two dimensions provides evidence that the family environment has a powerful impact on how students perceive university. In terms of returns at the labor market, differences in job enjoyment expectations close the gap by 11%, while wage expectations explain 13% of the SES gap in enrollment. While the origins of differences in beliefs about labor market returns could potentially be traced back to the observation of parental outcomes or outcomes in the neighborhood, it is clear that these belief gaps can account for a substantial proportion of the gap in enrollment decisions.

Next we investigate whether some of the SES gap in enrollment can be accounted for by differences in students' beliefs about their skills. Indeed we find that differences in perceived capabilities also play a key role accounting for 14% of the gap. While differences in perceived ability account for 11%, differences in the expected probability of getting good grades explain 4% of the gap.

We note that there may be other pecuniary and/or non-pecuniary factors we are not capturing with our survey which could also play a role in students' decisions to go to university. On the one hand, we cannot rule out that beliefs about omitted factors, which are correlated with the beliefs we elicit, might be biasing some individual factors upwards. On the other hand, the fact that we are not capturing all potentially relevant perceived returns associated with the respective decisions suggests that in the aggregate our estimates can be regarded as a lower bound.

Table 11: Share of SES enrollment gap closed when assigning beliefs of high to low SES

	Gap closed	95% CI
All returns	0.251	[0.194, 0.309]
Non-pecuniary returns	0.134	[0.083, 0.19]
Pecuniary returns	0.154	[0.125, 0.184]
Non-pecuniary returns at uni	0.089	[0.073, 0.104]
Social factors $q_1$	-0.036	[-0.039, -0.033]
Subject interest $q_2$	0.068	[0.062, 0.074]
Stress $q_3$	-0.005	[-0.012, 0.0]
Parental support $q_4$	0.06	[0.048, 0.071]
Life partner $q_5$	0.003	[0.001, 0.004]
Financial struggles	0.021	[0.016, 0.026]
Labor returns	0.211	[0.155, 0.265]
Job enjoyment	0.108	[0.057, 0.155]
Wages	0.133	[0.103, 0.164]
Capabilities	0.144	[0.121, 0.167]
Ability	0.105	[0.087, 0.123]
Getting good grades	0.042	[0.027, 0.057]

Notes: Each row stands for an experiment in which we randomly assign beliefs to low SES students drawn from the high SES belief distribution. The far left column specifies which beliefs are altered. ‘Gap closed’ indicates the share of SES gap accounted for. 95% confidence intervals are computed by bootstrapping the sample and drawing beliefs across 1000 simulations.

### 7.3 SES Gap - Differences in Preferences

To understand the role of differences in preferences across socio-economic groups, we assign preference parameters of high SES students to low SES students and investigate how this closes the gap in students’ enrollment decisions. In Table 12 we present the share of the SES gap accounted for by these experiments and the corresponding 95% confidence intervals.

We see that preference parameters for the non-pecuniary returns at university explain 2.8% of the gap. We find that despite having detailed information on perceived returns, unexplained preference differences for enrollment correlated with SES still account for 49% of the gap. The extracted utility from completion also plays a role contributing 7.6% of the SES gap. Taking all utility intercepts, i.e. unexplained prefer-

ences for each decision correlated with SES, together, we find that nearly two thirds of the SES gap is closed.

Table 12: Share of SES gap explained by preference differences

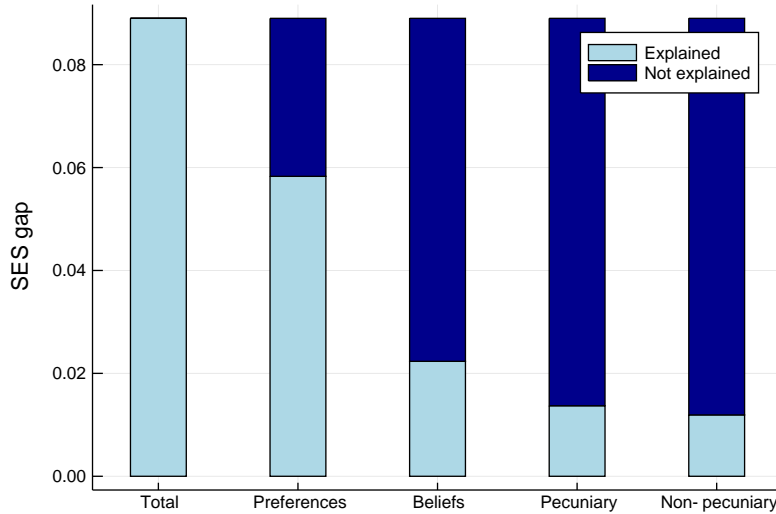
Parameter	Gap closed	95% CI
Non-pecuniary returns at uni	0.028	[0.021, 0.035]
Labor market returns	-0.013	[-0.015, -0.01]
Utility from enrollment	0.493	[0.461, 0.533]
Utility from completion	0.076	[0.07, 0.083]
Utility from enrollment and completion	0.561	[0.525, 0.602]
Utility from work (no degree)	0.115	[0.104, 0.127]
Utility from work (degree with low grades)	0.026	[0.023, 0.028]
Utility from work (degree with high grades)	-0.029	[-0.032, -0.027]
Utility from work (all three)	0.112	[0.102, 0.124]
All utility intercepts	0.655	[0.608, 0.708]

Notes: The table shows the share of the SES gap explained by assigning preferences of high SES students to low SES students. 95% confidence intervals are computed by bootstrapping the sample across 1000 simulations.

## 7.4 SES Gap - Summary

In Figure 5 we present the aggregate SES gap in the data vs. the heterogeneous preference model in the first bar. We show that the dynamic choice model (blue bar) is able to generate the entire gap observed in the data. In the second bar we present how much of the model gap can be accounted for by differences in preferences by SES (65%). The third bar shows that beliefs about pecuniary and non-pecuniary returns are also able to close a significant proportion of the model gap in enrollment (25%). The fourth bar exhibits the share of the gap explained by beliefs about pecuniary returns (15%), while the last bar shows the gap explained by beliefs about non-pecuniary returns (13%). The latter include beliefs about the immediate benefits/costs of attending university itself as well as beliefs about job enjoyment.

Figure 5: Decomposition of SES gap in enrollment decision



Notes: Bar 1 displays the total SES gap in the data and how much of it can be explained by our model (100%); bar 2 displays the fraction of the predicted gap that can be explained by differences in preferences across SES (65%); bar 3 displays the fraction of the predicted gap that can be explained by differences in beliefs about pecuniary and non-pecuniary returns (25%); bars 4 and 5 show the fraction of the predicted gap that can be explained by pecuniary (15%) and non-pecuniary factors (13%), respectively.

## 8 Discussion

Our study provides insights into which perceived returns play a role for university enrollment in general, and the SES gap in particular. Two questions that emerge are addressed in the following. First, are perceived returns comparable to observed university earnings premia (Section 8.1) and differences in university experiences (Section 8.2)? Second, where could these beliefs come from? In order to gain some understanding of the potential origins, we correlate beliefs with further background characteristics in Section 8.3.

As a disclaimer, we would like to note that we cannot make claims about whether beliefs about monetary and non-monetary returns are correct out of multiple reasons, one being that the observed returns are not causal due to self-selection. Even if we had credible causal estimates which we could compare to perceived returns, students in our sample might possess private information about their returns which we cannot observe,

further complicating our judgement about students' predictive accuracy. With these caveats in mind we compare perceived returns to descriptive differences.

## 8.1 Differences in realized earnings premia

In order to compare perceived monetary returns to college earnings premia observed in the data, we use the Labour Force Survey (LFS) data for individuals aged 27-33 in 2015. More specifically, we estimate college earnings premia and the returns to high grades while allowing the returns to differ by SES.<sup>24</sup> The results are reported in columns 1 and 3 of Table Table 13. We compare these estimates to students' perceptions about returns which we report in columns 2 and 4 of the same table. In order to compute these columns, we take each student as two observations. For column 2 we once take the observation of the student with perceived earnings without going to university as a dependent variable and once with expected earnings from graduating from university, i.e. the convex combination of perceived earnings with low and high grades weighted by the perceived probability of getting high grades. For column 4 we take one observation of each student's perceived earnings with low grades and another with high grades.

In the first column of Table 13 we show that the average earnings premium for full-time high SES workers aged around 30 years in the UK is 42 log points. Computing the same return in our sample, in the second column we find that the average perceived return is 49 log points. The interaction between university and low SES in the data is insignificant at -8 log points, whereas the perceived low SES penalty is 12 log points and significant. The coefficient on the low SES dummy is -9 log points both for observed and perceived returns.

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<sup>24</sup>In the LFS, low and high SES families are defined as those for whom the main wage earner falls into categories 1-3 and categories 4-9 of the ONS standard occupation classification, respectively.

Table 13: Comparison between observed and perceived returns to university and good grades in terms of log earnings

	Return to uni		Return to good grades	
	LFS	Beliefs	LFS	Beliefs
University degree	0.42*** (0.04)	0.49*** (0.03)		
Low SES x Uni	-0.08 (0.05)	-0.12*** (0.04)		
Good grades			0.23*** (0.05)	0.21*** (0.03)
Low SES x Good grades			-0.07 (0.08)	-0.06 (0.04)
Female	-0.19*** (0.03)	-0.28*** (0.02)	-0.09** (0.04)	-0.30*** (0.02)
Low SES	-0.09** (0.04)	-0.09*** (0.03)	-0.09 (0.07)	-0.16*** (0.03)
Constant	10.10*** (0.03)	10.05*** (0.03)	10.29*** (0.05)	10.38*** (0.03)
R-squared	0.251	0.110	0.208	0.062
Observations	1105	4750	408	4750

*Notes:* The first and third column are estimated using the 2015 Labour Force Survey while restricting the sample to individuals aged 27-33 and controlling for age distance from 30. The second and fourth column are estimated using our sample while taking each individual twice, once for each scenario. The first two columns estimate the difference between the log of earnings of the full sample in order to capture the college premium, while the last two columns focus on the return to good grades conditional on having graduated from university. Estimation technique is OLS. Standard errors in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

We repeat the same exercise for the observed and perceived returns to good grades. For high SES the earnings gap between lower and higher grades is 23 log points, compared to perceived 21 log points in our sample. The interaction between good grades and low SES is an insignificant negative 7 and negative 6 log points in the LFS and our sample, respectively. It is remarkable that, on average, low and high SES students have beliefs about earnings premia and levels which line up relatively closely with what is observed in the data.



## 8.2 Differences in actual university experiences

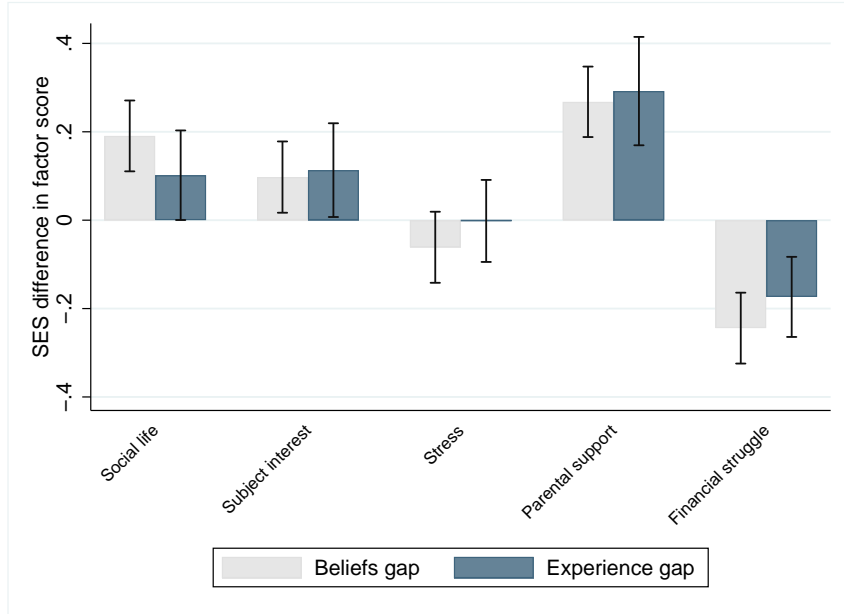
We document large and systematic SES gaps in student beliefs about different immediate benefits and costs of attending university. A natural question to ask is whether the immediate returns to attending university actually differ between low and high SES students. While we cannot observe students' outcomes in the counterfactual state in which they choose a different education path, we shed some light on this question by studying a representative sample of 1,002 university students in England. The analysis is presented in more detail in Boneva, Golin and Rauh (2019) and reveals striking differences between first- and continuing-generation university students' self-reported current experiences.<sup>25</sup>

Compared to first-generation students, continuing-generation students enjoy their social life more and they find the subject they are studying more interesting. There is no difference across these two groups in terms of how stressful the students perceive studying to be. Continuing-generation students further report higher parental approval in their choice to go to university and they are less likely to report that they struggle financially. As illustrated in Figure 6 it is remarkable how closely these average differences in actual experiences line up with the differences in beliefs that we measure. Boneva, Golin and Rauh (2019) further find that first-generation students are less likely to agree with the statement that life at university is better than expected. The time use data reported in that study also reveals that low SES students spend more time working for pay alongside their studies and less time socializing with friends. While this evidence is solely indicative, it does suggest that actual differences exist in terms of how students from different backgrounds experience university life. Further research will be needed to better understand what might be driving these differences and which policies may narrow these socio-economic gaps.

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<sup>25</sup>Other recent studies investigating gaps in university experiences across students of different background include Janke et al. (2017) and Delavande, Del Bono and Holford (2019).

Figure 6: Perceived vs. Actual Gaps in University Experiences



Notes: The grey bars illustrate the socio-economic differences in secondary school students' beliefs about the immediate benefits/costs of university education calculated from the data presented in this paper. The blue bars use the data presented in Boneva, Golin and Rauh (2019) and illustrate the gap in actual experiences of university students. The black lines represent 95% confidence intervals. See Boneva, Golin and Rauh (2019) for a complete description of the data.

### 8.3 Determinants of beliefs about returns

To gain a better understanding of potential determinants of perceived returns, we look at how individual and school neighborhood characteristics correlate with perceived returns. Appendix Tables A.7 and A.8 show how these characteristics are related to students' beliefs about the immediate and later-life returns to university, respectively. As described in detail above, students from a high SES background perceive the benefits of university attendance as higher and the immediate costs as lower. Interestingly, students who report that they have an older friend or sibling who has been to university and students who report a higher number of people they can ask about university life perceive the benefits in terms of their social life to be higher and the costs in terms of stress and financial struggles to be lower.

We also find that the number of people the student can ask about university is

positively related to students' perceptions about the labor market returns in terms of earnings and job enjoyment. Finally, students who go to school in areas where a higher proportion of adults have university degrees perceive the returns in terms of subject interest, parental support and earnings to be higher, while they perceive the costs in terms of financial struggles to be lower. Given that these variables are endogenous and may in part be determined by students' socio-economic background, we cannot interpret this evidence as causal. The results are, however, consistent with a model in which exposure to people who have experienced university life positively affects students' views about their likely university experience and the monetary returns to university education.

## 9 Conclusion

In this study, we elicit beliefs about different pecuniary and non-pecuniary benefits and costs of university education. We first document that there is a significant amount of heterogeneity in these beliefs and that there are significant differences in perceptions about returns across socio-economic groups. We build and estimate a dynamic choice model, in which we allow students to decide whether to study, work alongside their studies, complete their degree and work once they have completed their education. We model students' decisions to work alongside their studies as a function of their financial situation, and allow different immediate non-pecuniary benefits and costs of university education to differ depending on whether the student decides to work alongside university or not.

Overall, the dynamic model can explain the entire SES gap in university enrollment. Differences in beliefs about pecuniary and non-pecuniary returns can account for 25% of the predicted gap, while 13% of the predicted gap can be explained by differences in beliefs about non-pecuniary returns alone. Amongst the non-pecuniary factors, we find that students' interest in the subject/material, students' beliefs about whether their parents will support them in their choice, and students' beliefs about the likelihood

of enjoying the job they will be doing are particularly important, both in students' enrollment decisions as well as in explaining the socio-economic gap.

Students' beliefs about pecuniary returns seem to be remarkably similar to descriptive college-earnings premia. Moreover, SES gaps in students' beliefs about immediate non-pecuniary returns seem to line up well with actual SES gaps in university experiences. More research will be needed into why the actual experiences of students from different backgrounds differ and which policies can be put in place to mitigate the SES gap in experiences as well as the SES gap in enrollment.

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# Online Appendix

## A Supplementary Analyses

Table A.1: University enrollment gap

Dependent variable: Attended further education (dummy)			
	(1)	(2)	(3)
Female	0.039*** (0.01)	0.091*** (0.01)	0.077*** (0.01)
University (parent)	0.189*** (0.01)	0.138*** (0.02)	0.152*** (0.01)
Controls	No	Yes	Yes
Cohort dummy	No	No	Yes
R-Squared	0.03	0.08	0.23
N	7173	5153	5153

Datasource: BHPS, UKHLS.

Notes: Estimation technique is OLS. Standard errors in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . The sample includes all respondents born after 1980 that are surveyed at least once after the age of 18 residing in England. All regressions include a constant. *University (parent)* is a dummy taking the value one if at least one parent has a university degree. Controls include a constant, scores from each of the Big Five personality traits (openness, agreeableness, conscientiousness, extraversion, neuroticism) and from both cognitive and verbal tests. The *Cohort dummy* includes a dummy for each year of birth.

Table A.2: Summary statistics

Variable	Mean	Std. Dev.	Min	Max	N
Male respondent	0.371	0.483	0	1	2540
School year	11.223	1.570	9	13	2540
At least one parent has degree	0.553	0.497	0	1	2411
People to ask about university	5.957	3.792	0	11	2515
Older sibling/friend university	0.611	0.488	0	1	2351

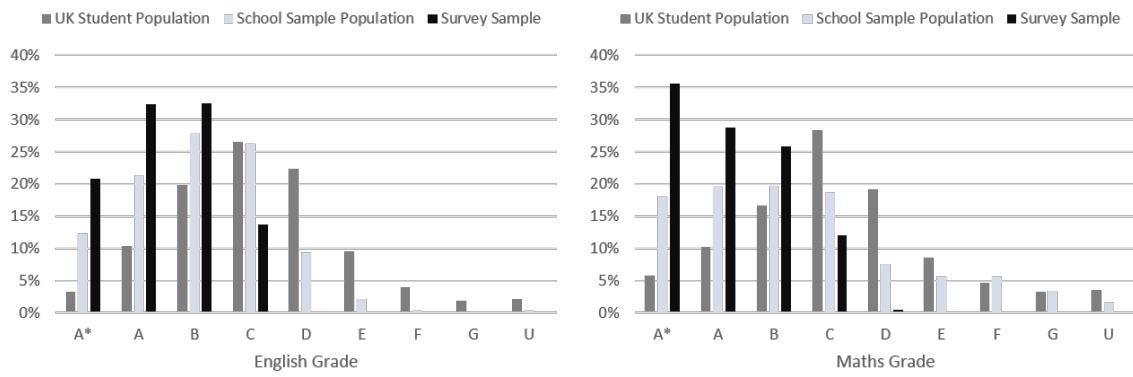
Notes: 'People to ask about university' refers to the stated number of people the respondent knows whom he/she can ask about university life.

Figure A.1: Location of schools in sample



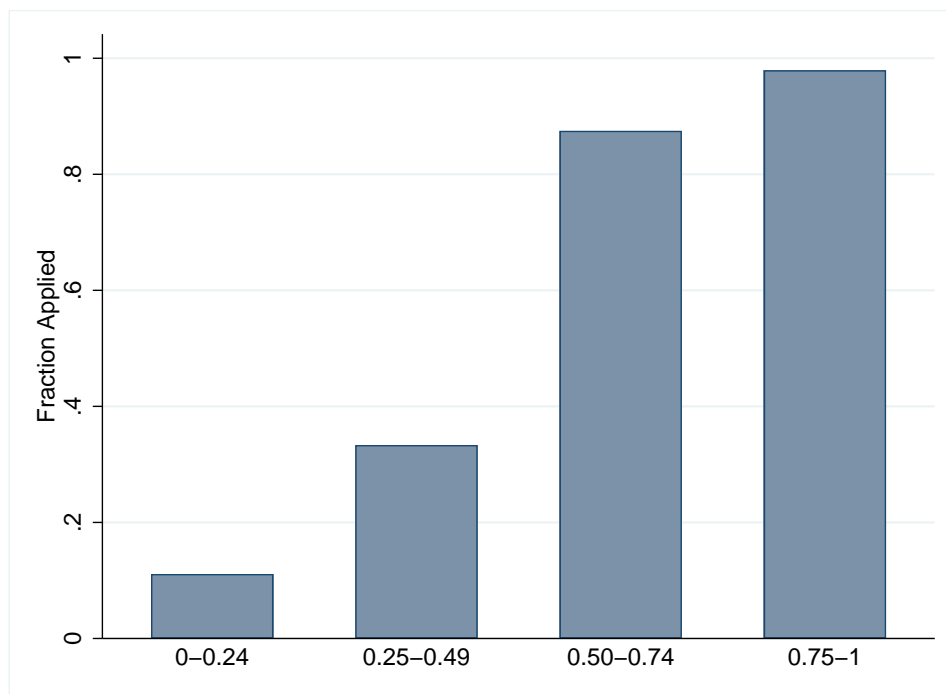
Note: This map shows the location of the schools in our sample. The schools which are marked in blue offer a sixth form (i.e. Years 12 and 13) while the schools marked in yellow do not offer a sixth form.

Figure A.2: Distribution of GCSE grades



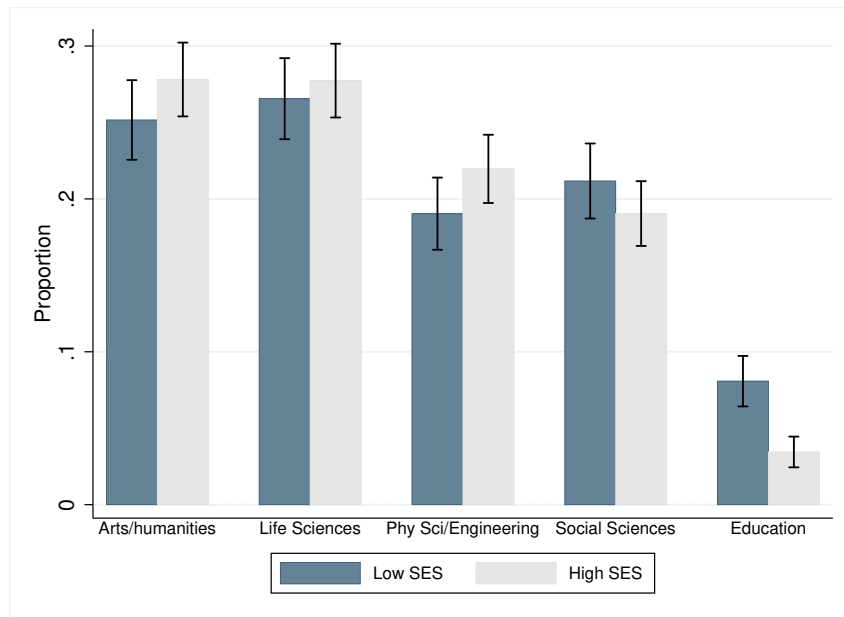
Note: The two panels display the distribution of GCSE English Literature and Mathematics grades, respectively, for (i) the UK student population, (ii) the population of students in the surveyed schools and (iii) the survey sample.

Figure A.3: Fraction applied by stated intentions



Notes: This figure shows the fraction of students in their final year who applied to university conditional on their stated intentions to apply. Stated intentions are summarized in four bins: 0-0.24, 0.25-0.49, 0.50-0.74, 0.75-1.

Figure A.4: Intended university subject field by socio-economic status



Note: This graph shows the proportion of respondents intending to study each of the five specified subject fields, separately by socio-economic status. 95% confidence intervals are provided.

Table A.3: Spearman rank correlations of university beliefs

	Enjoy social	Meet people	Lose contact	Lonely	Interest	Enjoy	Hard	Stress	Parental support	Life partner	Struggle financially	Enough money	Financial support
Enjoy social	1.000												
Meet people	0.593	1.000											
Lose contact	-0.067	-0.108	1.000										
Lonely	-0.299	-0.376	0.342	1.000									
Interest	0.309	0.323	-0.054	-0.056	1.000								
Enjoy	0.330	0.316	-0.086	-0.141	0.647	1.000							
Hard	-0.050	-0.103	0.197	0.298	-0.130	-0.181	1.000						
Stress	-0.122	-0.172	0.198	0.418	-0.097	-0.188	0.580	1.000					
Parental support	0.219	0.216	-0.112	-0.099	0.280	0.235	-0.036	-0.012	1.000				
Life partner	0.139	0.197	-0.009	-0.092	0.089	0.094	-0.024	-0.072	0.047	1.000			
Struggle financially	-0.039	-0.078	0.233	0.289	-0.036	-0.121	0.400	0.443	-0.042	-0.097	1.000		
Enough money	0.160	0.192	-0.110	-0.153	0.154	0.217	-0.149	-0.179	0.136	0.191	-0.426	1.000	
Financial support	0.115	0.115	-0.086	-0.038	0.110	0.098	0.064	0.056	0.263	0.103	-0.081	0.299	1.000

Notes: This table shows Spearman rank correlations for beliefs about immediate outcomes in the scenario in which students are asked to imagine that they go to university. The sample size is 2,540 students. See Table 1 for a full list of questions.



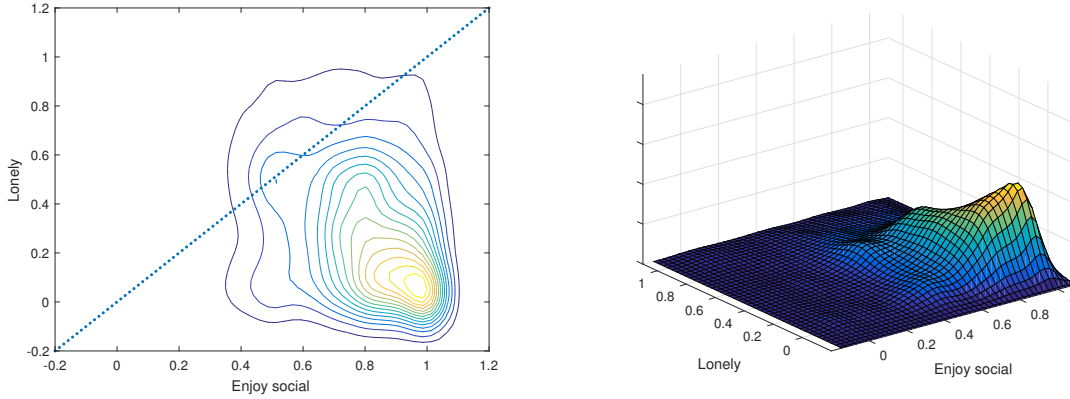
Table A.4: Spearman rank correlations of work beliefs

	Enjoy social	Meet people	Lose contact	Lonely	Interest	Enjoy	Hard	Stress	Parental support	Life partner	Struggle financially	Enough money	Financial support
Enjoy social	1.000												
Meet people	0.628	1.000											
Lose contact	-0.122	-0.103	1.000										
Lonely	-0.306	-0.353	0.353	1.000									
Interest	0.469	0.469	-0.043	-0.238	1.000								
Enjoy	0.456	0.467	-0.057	-0.219	0.762	1.000							
Hard	-0.010	0.006	0.235	0.205	0.129	0.087	1.000						
Stress	-0.110	-0.083	0.250	0.342	0.033	0.004	0.630	1.000					
Parental support	0.251	0.239	-0.109	-0.149	0.295	0.275	0.087	0.033	1.000				
Life partner	0.268	0.293	0.026	-0.142	0.270	0.241	0.063	-0.007	0.146	1.000			
Struggle financially	-0.092	-0.079	0.272	0.290	-0.039	-0.081	0.374	0.455	-0.048	0.003	1.000		
Enough money	0.280	0.258	-0.103	-0.166	0.253	0.280	-0.038	-0.133	0.207	0.177	-0.398	1.000	
Financial support	0.118	0.125	0.009	-0.024	0.202	0.163	0.184	0.157	0.246	0.137	0.198	0.085	1.000

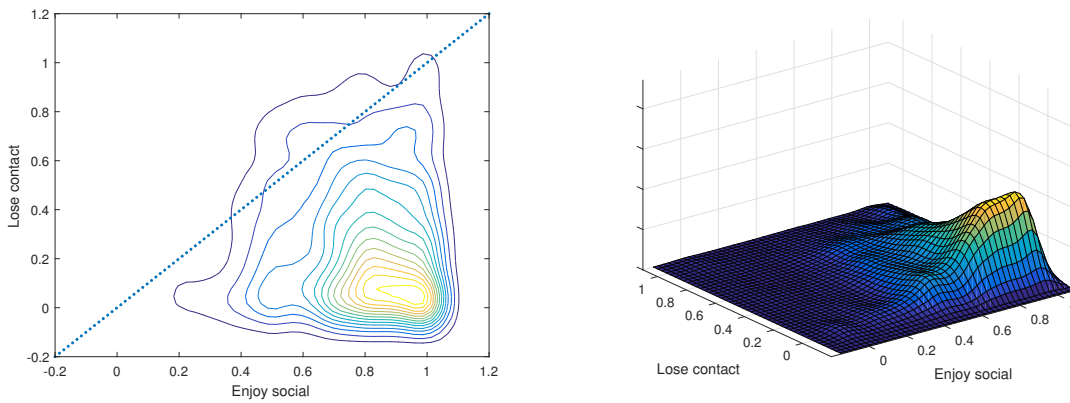
*Notes:* This table shows Spearman rank correlations for beliefs about immediate outcomes in the scenario in which students are asked to imagine that they do not go to university but start working instead. The sample size is 2,540 students. See Table 1 for a full list of questions.

Figure A.5: Distribution of beliefs

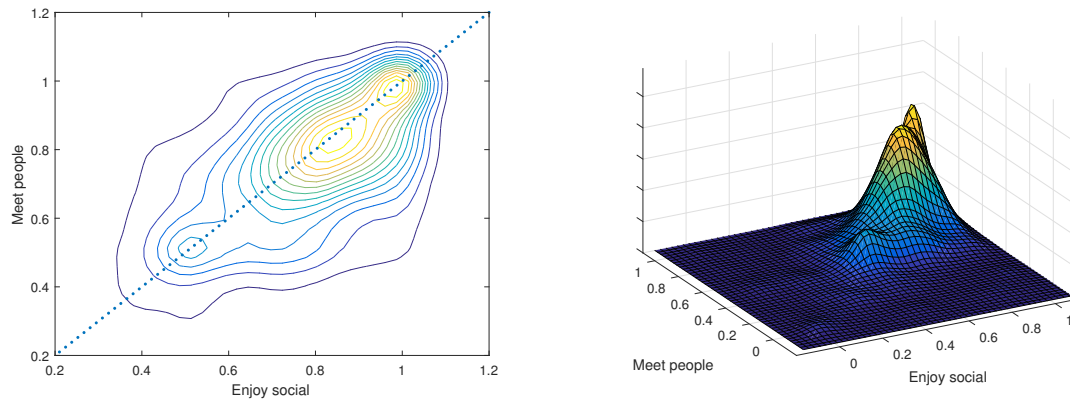
A: Enjoy social life and feel lonely



B: Enjoy social life and lose contact

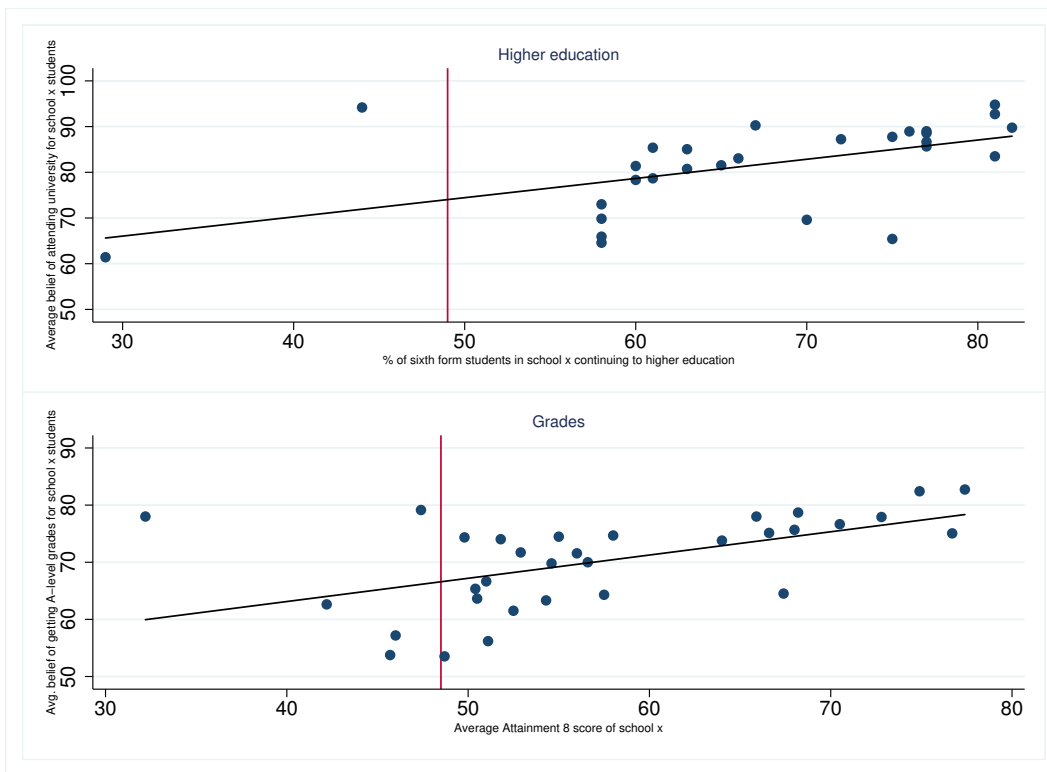


C: Enjoy social life and meet people



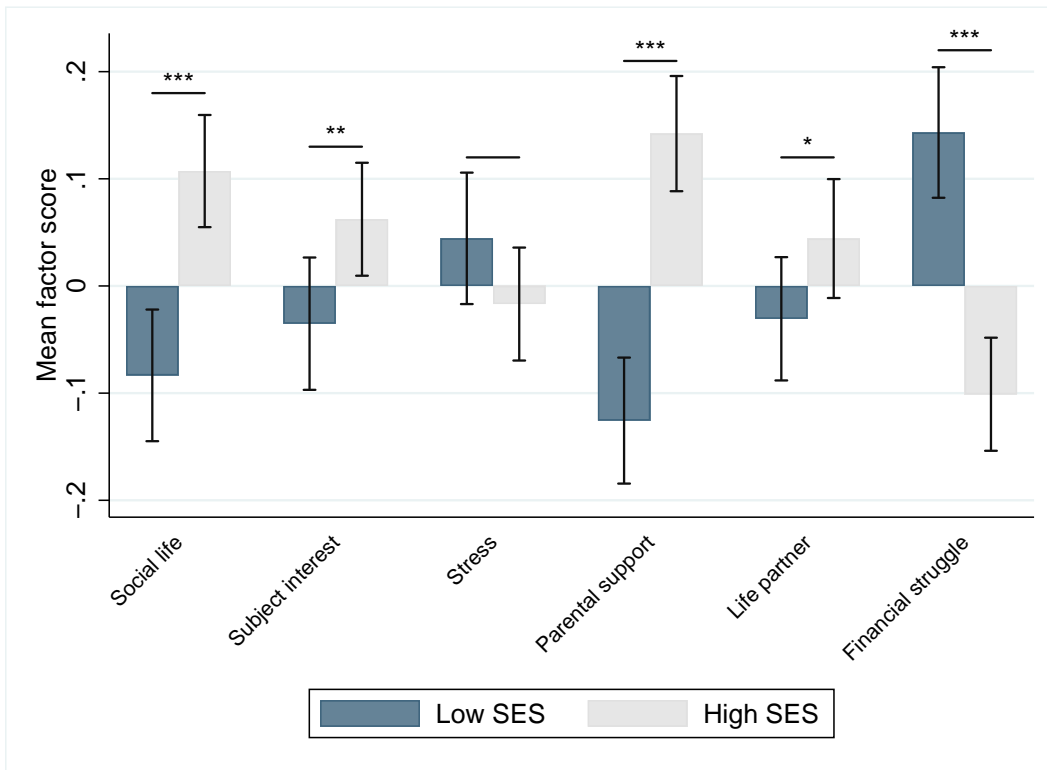
Notes: This figure shows the joint distributions of the perceived likelihood of two social outcomes conditional on going to university. The left panels give the contour plots, while the right panels show density distributions. Panel A looks at the joint distribution of the belief that you will enjoy the social life and feel lonely conditional on going to university, while Panel B looks at the likelihood of enjoying the social life and losing contact with family and friends. Panel C shows the joint distribution for the perceived likelihood of enjoying the social life and meeting people.

Figure A.6: Mean beliefs and school level data



Notes: These graphs plot average school level beliefs against actual school level averages. The vertical red line indicates the national average.

Figure A.7: Mean belief factor scores by SES



Notes: The figure shows average factor scores for the immediate returns to university separately by socio-economic status. High SES students are defined as those students who have at least one parent with university education. Stars indicate differences by SES: \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

Figure A.8: Perceived earnings and probability of employment by SES



Notes: Panel A shows average perceived earnings (cond. on having a job) while Panel B shows the average perceived probability of having a full-time job separately for the scenarios: (i) work rather than go to university, (ii) go to university and graduate with low grades, and (iii) go to university and graduate with high grades. High SES students are defined as those students who have at least one parent with university education. Stars indicate differences by SES: \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

Table A.5: Differences in beliefs about immediate outcomes

Belief	Low SES				High SES				Diff-in-diff	Diff-in-var
	University	Work	Difference	N	University	Work	Difference	N		P-value
Enjoy social life and activities	74.91 [21.49]	63.77 [24.50]	11.14*** [26.62]	1,077	77.19 [20.56]	63.04 [24.03]	14.14*** [25.05]	1,334	3.01*** (1.06)	0.035
Meet people	71.69 [21.67]	61.47 [23.88]	10.22*** [24.71]	1,077	74.84 [19.75]	61.68 [22.88]	13.15*** [23.58]	1,334	2.94*** (0.99)	0.107
Lose contact with family/friends	27.98 [26.97]	21.26 [24.35]	6.72*** [24.71]	1,077	28.64 [25.85]	24.69 [24.74]	3.95*** [23.83]	1,334	-2.77*** (0.99)	0.208
Feel lonely	32.37 [26.17]	35.69 [27.92]	-3.33*** [27.06]	1,077	30.57 [25.77]	38.99 [28.69]	-8.42*** [26.81]	1,334	-5.09*** (1.10)	0.743
Find material/work tasks interesting	69.13 [21.88]	53.99 [25.92]	15.14*** [32.64]	1,077	73.49 [20.36]	55.03 [25.43]	18.46*** [30.68]	1,334	3.31** (1.29)	0.032
Enjoy studying/work	69.61 [22.15]	56.90 [25.33]	12.71*** [29.72]	1,077	72.06 [21.55]	57.23 [24.29]	14.82*** [28.83]	1,334	2.12* (1.20)	0.292
Find material hard/workload high	54.12 [23.81]	42.64 [24.34]	11.48*** [27.60]	1,077	52.31 [23.97]	41.44 [24.93]	10.88*** [26.69]	1,334	-0.60 (1.11)	0.246
Feel stressed	54.68 [28.18]	42.59 [28.09]	12.09*** [29.52]	1,077	51.73 [29.17]	42.09 [28.09]	9.64*** [28.32]	1,334	-2.45** (1.18)	0.149
Have parental support in choice	82.55 [25.31]	62.92 [33.91]	19.63*** [36.92]	1,077	88.45 [20.55]	58.76 [34.73]	29.69*** [37.58]	1,334	10.06*** (1.53)	0.543
Meet life partner	42.63 [26.06]	38.48 [26.46]	4.15*** [21.94]	1,077	46.20 [25.49]	40.34 [25.95]	5.85*** [23.56]	1,334	1.71* (0.94)	0.014
Struggle financially	56.24 [29.11]	37.63 [28.11]	18.61*** [33.88]	1,077	50.44 [29.95]	38.19 [27.20]	12.25*** [32.68]	1,334	-6.37*** (1.36)	0.212
Have enough money	46.49 [25.23]	56.94 [25.67]	-10.45*** [29.25]	1,077	54.18 [24.79]	56.08 [24.96]	-1.90** [28.90]	1,334	8.54*** (1.19)	0.677
Have financial support from parents	60.12 [31.32]	45.94 [33.14]	14.17*** [29.45]	1,077	67.33 [27.73]	50.33 [31.43]	17.00*** [29.76]	1,334	2.82** (1.21)	0.721

*Notes:* Standard deviations given in square brackets, and standard errors given in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . This table provides mean beliefs about immediate outcomes separately by SES. Columns 1-4 are for respondents for whom neither parent went to university, while Columns 5-8 are for respondents for whom at least one parent went to university. Within each group, the first two columns give mean beliefs for the respective aspect under the scenarios of going to university and not going to university, respectively. Mean beliefs are given on a 0-100 scale. The third column gives the mean difference between these two beliefs. T-tests are used to test for equality of means. Column 9 ('Diff-in-diff') gives the average difference for respondents for whom at least one parent has a degree minus the average difference for respondents for whom neither parent has a degree. Column 10 ('Diff-in-var') gives the p-value for a test of equality of variances between the perceived returns.

Table A.6: Differences in beliefs about later-life outcomes

	All	SES		Difference (p-value)
		Low SES	High SES	
<i>A: No degree</i>				
Earnings	23,912 (18,379)	22,966 (17,259)	24,867 (18,609)	1,902** (0.010)
Employed	0.687 (0.259)	0.715 (0.255)	0.671 (0.257)	-0.044*** (0.000)
Enjoy job	0.518 (0.256)	0.528 (0.260)	0.510 (0.250)	-0.019* (0.075)
<i>B: Degree with low grades</i>				
Earnings	31,256 (22,156)	29,499 (21,370)	33,051 (22,212)	3,552*** (0.000)
Employed	0.771 (0.213)	0.780 (0.217)	0.772 (0.203)	-0.008 (0.334)
Enjoy job	0.666 (0.215)	0.672 (0.222)	0.665 (0.207)	-0.007 (0.411)
<i>C: Degree with high grades</i>				
Earnings	37,450 (24,995)	34,144 (23,756)	40,410 (25,028)	6,266*** (0.000)
Employed	0.856 (0.175)	0.860 (0.177)	0.862 (0.163)	0.002 (0.792)
Enjoy job	0.730 (0.192)	0.730 (0.196)	0.734 (0.184)	0.003 (0.672)
<i>D: Difference (low grades - no degree)</i>				
Earnings	7,344 (16,640)	6,533 (15,411)	8,187 (17,265)	1,651** (0.014)
Employed	0.084 (0.230)	0.065 (0.218)	0.101 (0.232)	0.035*** (0.000)
Enjoy job	0.148 (0.247)	0.144 (0.244)	0.155 (0.245)	0.011 (0.257)
<i>E: Difference (high grades - low grades)</i>				
Earnings	6,194 (15,726)	4,645 (14,390)	7,359 (16,275)	2,714*** (0.000)
Employed	0.084 (0.176)	0.080 (0.172)	0.090 (0.175)	0.010 (0.154)
Enjoy job	0.064 (0.196)	0.059 (0.194)	0.069 (0.196)	0.010 (0.190)

*Notes:* Column 1 presents the mean across all respondents while columns 2 and 3 show the means for low SES and high SES students, respectively. Column 4 displays the difference in means along with the corresponding p-value testing differences in means. High SES is defined as having at least one parent who has a university degree. Panels A, B and C present average responses for earnings conditional on working full-time, the stated likelihood of being employed and enjoying the job one will be doing in the scenarios in which one does not obtain a university degree (A), obtains a degree with a low grade (B) and obtains a degree with a high grade (C). Panels D and E present the means in differences in responses across scenarios as well as socio-economic differences in perceived differences. \* p<0.10, \*\* p<0.05, \*\*\* p<0.01.

Table A.7: Determinants of beliefs about immediate returns

	(1)	(2)	(3)	(4)	(5)	(6)
	Social life	Subject interest	Stress	Parental support	Life partner	Financial struggle
High SES	0.156*** (0.04)	0.142*** (0.04)	0.002 (0.04)	0.175*** (0.04)	0.075* (0.04)	-0.104** (0.04)
Older sibling/friend university	0.135*** (0.04)	-0.030 (0.04)	-0.033 (0.04)	0.050 (0.04)	-0.006 (0.05)	-0.157*** (0.04)
People to ask about university	0.019*** (0.01)	0.001 (0.01)	-0.011* (0.01)	0.009 (0.01)	0.008 (0.01)	-0.011* (0.01)
Male student	-0.022 (0.04)	-0.002 (0.04)	-0.138*** (0.04)	0.014 (0.04)	-0.005 (0.04)	-0.057 (0.04)
% University degree (area)	0.002 (0.00)	0.003* (0.00)	0.000 (0.00)	0.017*** (0.00)	0.000 (0.00)	-0.008*** (0.00)
School year FEs	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.05	0.06	0.03	0.07	0.01	0.08
N	2347	2347	2347	2347	2347	2347

*Notes:* Estimation technique is OLS. Standard errors in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . The independent variables are the six extracted factors summarizing perceived immediate returns to university. High SES students are defined as those students who have at least one parent with a university degree. The regression additionally controls for a constant, whether or not the student has an older friend or sibling who has gone to university, the number of people the student can think of he/she can ask about university life, a male dummy, the percent of adults with a university degree in the local area and school year fixed effects.



Table A.8: Determinants of beliefs about later-life returns

	Low grades - no uni		High grades - low grades	
	(1) Earnings	(2) Enjoy job	(3) Earnings	(4) Enjoy job
High SES	421.977 (740.43)	0.012 (0.01)	1844.040*** (644.65)	0.001 (0.01)
Older sibling/friend university	-880.404 (731.71)	-0.001 (0.01)	1035.210 (695.97)	0.004 (0.01)
People to ask about university	370.743*** (101.28)	-0.001 (0.00)	130.666 (94.47)	0.003** (0.00)
Male student	3004.702*** (753.35)	0.000 (0.01)	1328.603* (682.41)	-0.013 (0.01)
% University degree (area)	70.629** (31.30)	0.001 (0.00)	44.436 (28.24)	0.001 (0.00)
School year FEs	Yes	Yes	Yes	Yes
R-squared	0.02	0.00	0.02	0.01
N	2347	2347	2347	2347

*Notes:* Estimation technique is OLS. Standard errors in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . The independent variables are perceived later-life returns in terms of earnings and job enjoyment. Columns 1 and 2 display perceived returns to graduating with low grades versus not having a university degree while columns 3 and 4 show perceived returns to graduating with high grades versus graduating with low grades. High SES students are defined as those students who have at least one parent with a university degree. The regression additionally controls for a constant, whether or not the student has an older friend or sibling who has gone to university, the number of people the student can think of he/she can ask about university life, a male dummy, the percent of adults with a university degree in the local area and school year fixed effects.

## B Questionnaires

### B.1 Plans for the future

1. *How likely do you think it is that you will obtain the required A-level grades to go to university?* [0-100 scale where 0 means very unlikely and 100 means very likely]
2. *Assuming you do get the grades in sixth form/college to go to university, how likely do you think it is that you will go to university?* [0-100 scale]
3. *Which field of study would you be most likely to choose if you decided to go to university? We know this is a difficult question. Please select one field of study which you think interests you most, even if you are unsure about which field of study best fits your interests*
  - (a) Arts and Humanities (e.g. languages, history, music, architecture, philosophy)
  - (b) Life Sciences (e.g. biology, medicine, pharmacy, psychology)
  - (c) Physical Sciences and Engineering (e.g. mathematics, computer science, physics, engineering)
  - (d) Social Sciences (e.g. economics, law, business)
  - (e) Education

*Now imagine you have been admitted to university and you started an undergraduate degree in the field of [insert chosen subject from Question 3 above].*

1. *How likely do you think it is that you would complete your degree?* [0-100 scale]
2. *Assuming that you graduate, how likely do you think it is that you would get a First or a 2.1?* [0-100 scale]

## B.2 Hypothetical Scenarios

*Now we would like you to think about the 3-4 years of your life that will come after you leave sixth form/college. Imagine that during those 3-4 years you go to university. What do you think your life during those 3-4 years will be like?*

*If you go to university, how likely do you think it is that you will... [0-100 scale]*

1. ...enjoy the social life and activities you engage in?
2. ...meet people with whom you easily get along with?
3. ...lose contact with your family and current friends?
4. ...feel lonely and not part of a group?
5. ...find the material that is covered interesting?
6. ...enjoy studying for the course?
7. ...find the material too hard and/or work load too high?
8. ...be stressed and anxious about not being able to cope?
9. ...struggle financially?
10. ...receive financial support from your family?
11. ...have enough money to do what you enjoy?
12. ...have support from your parents in your decision to go to university rather than work immediately?
13. ...meet your future partner
14. ...have to work alongside your studies?

*Please keep thinking about the 3-4 years of your life that will come after you leave sixth form/college. This time imagine that during those 3-4 years you do not go to university but start working instead. What do you think your life during those 3-4 years will be like?*

*If you start working, how likely do you think it is that you will... [0-100 scale]*

1. ...enjoy the social life and activities you engage in?
2. ...meet people with whom you easily get along with?
3. ...lose contact with your family and current friends?
4. ...feel lonely and not part of a group?
5. ...find the tasks at work interesting?
6. ...enjoy performing the different tasks at work?
7. ...find the tasks at work too hard and/or work load too high?
8. ...be stressed and anxious about not being able to cope?
9. ...struggle financially?
10. ...receive financial support from your family?
11. ...have enough money to do what you enjoy?
12. ...have support from your parents in your decision to work rather than attend university?
13. ...meet your future partner

*Now we would like you to think about what your life might be like when you are 30 years old. What do you think your life will be like if you did enroll in a [insert stated subject choice] degree and graduated with a First or a 2.1? Try to think about the types of jobs that would be available to you and answer the following questions.*

1. *How likely do you think it is that you will have a paid job at age 30? [0-100 scale]*
2. *Assuming that you work full-time, how likely do you think it is that you would enjoy the job that you would be doing? [0-100 scale]*
3. *Assuming that you work full-time and that there is no inflation, what do you think your pre-tax earnings are likely to be per year? [£0, £1,000, £2,000,... £99,000, £100,000, More than £100,000]*

*Please continue to think about what your life might be like when you are 30 years old.*

*What do you think your life will be like if you did enroll in a [insert stated subject choice] degree and graduated but NOT a First or a 2.1? Try to think about the types of jobs that would be available to you and answer the following questions.*

1. *How likely do you think it is that you will have a paid job at age 30? [0-100 scale]*
2. *Assuming that you work full-time, how likely do you think it is that you would enjoy the job that you would be doing? [0-100 scale]*
3. *Assuming that you work full-time and that there is no inflation, what do you think your pre-tax earnings are likely to be per year? [£0, £1,000, £2,000,... £99,000, £100,000, More than £100,000]*

*Please continue to think about what your life might be like when you are 30 years old.*

*What do you think your life will be like if you did not go to university or did go to university but did not graduate? Try to think about the types of jobs that would be available to you and answer the following questions.*

1. *How likely do you think it is that you will have a paid job at age 30? [0-100 scale]*
2. *Assuming that you work full-time, how likely do you think it is that you would enjoy the job that you would be doing? [0-100 scale]*
3. *Assuming that you work full-time and that there is no inflation, what do you think your pre-tax earnings are likely to be per year? [£0, £1,000, £2,000,... £99,000, £100,000, More than £100,000]*

### B.3 Follow-up survey (wave 2)

1. *How likely do you think it is that you will obtain the required A-level grades to go to university?* [0-100 scale]
2. *Assuming you do get the grades in sixth form/college to go to university, how likely do you think it is that you will go to university* [0-100 scale]
3. *Did you apply to university?* [Yes, No] (only asked to respondents in Year 13)
  - (a) *Which field of study did you choose?* (if answer was 'Yes')
    - i. Arts and Humanities (e.g. languages, history, music, architecture, philosophy)
    - ii. Life Sciences (e.g. biology, medicine, pharmacy, psychology)
    - iii. Physical Sciences and Engineering (e.g. mathematics, computer science, physics, engineering)
    - iv. Social Sciences (e.g. economics, law, business)
    - v. Education
    - vi. Other

## C The Dynamic Choice Problem

### C.1 Solving Choice Problem Backwards

Note that all choices are made ex-ante and therefore are not binary but are in terms of probabilities  $p_j$  for  $j \in \{s, c, l\}$  as individuals are not aware of their future shock realizations and can only anticipate them based on their distributions. Given the closed form solution of the difference of two extreme value distribution shocks, we can write down the solutions to the decisions of the model. At  $t = 3$  when the student decides whether to work or not for each of the three final decision nodes  $d$ , the student will work if:

$$\begin{aligned} \tau v(w_d) + \lambda \pi_d + U_{ld}(X_l) &\geq -\varepsilon_{ld} \\ \iff p_{ld} &= \frac{1}{1 + e^{-(\tau v(w_d) + \lambda \pi_d + U_{ld}(X_l))}} \end{aligned}$$

These three decision probabilities  $p_{l^0}$ ,  $p_{l^1}$ , and  $p_{l^2}$  will allow the student to compute the expected values from the final period. This will allow her to think about her decision at  $t = 1$ , in case she is enrolled in university, where she will complete university if

$$\begin{aligned} E[V_2^{c1}] + U_c(X_c) + \varepsilon_c &\geq E[V_2^{c0}] \\ \iff E[V_2^{c1}] - E[V_2^{c0}] + U_c(X_c) &\geq -\varepsilon_c \\ \iff p_c &= \frac{1}{1 + e^{-(E[V_2^{c1}] - E[V_2^{c0}] + U_c(X_c))}} \end{aligned}$$

Now knowing both  $p_c$  and  $p_{ld}$ , at  $t = 0$  the student is ready to decide between three options: (i) not enrolling, (ii) enrolling and working alongside university ( $s_1$ ), (iii) enrolling without working alongside ( $s_2$ ). The student will choose  $s_1$  if:

$$\begin{aligned} E[V_1^{s1}] + U_{s1}(X_s) + \varepsilon_{s1} &\geq E[V_1^{s0}] \wedge E[V_1^{s1}] + U_{s1}(X_s) + \varepsilon_{s1} \geq E[V_1^{s2}] \\ \iff E[V_1^{s1}] - E[V_1^{s0}] + U_{s1}(X_s) &\geq -\varepsilon_{s1} \wedge E[V_1^{s1}] - E[V_1^{s2}] + U_{s1}(X_s) \geq -\varepsilon_{s1} \end{aligned}$$

$$\Leftrightarrow p_{s_1} = \frac{e^{E[V_1^{s_1}] + U_{s_1}(X_s)}}{1 + e^{E[V_1^{s_1}] + U_{s_1}(X_s)} + e^{E[V_1^{s_2}] + U_{s_2}(X_s)}}$$

and analogously will choose  $s_2$  with:

$$p_{s_2} = \frac{e^{E[V_1^{s_2}] + U_{s_2}(X_s)}}{1 + e^{E[V_1^{s_1}] + U_{s_1}(X_s)} + e^{E[V_1^{s_2}] + U_{s_2}(X_s)}}.$$

## C.2 Estimation Details

In Table C.1 we present the variance-covariance matrix for the unobserved heterogeneity for low (top) and high (bottom) SES students. The variance for the unobserved heterogeneity of the work decision is normalized to unity. We see that for low SES the decision to enroll and not to work alongside studies has the next highest variance (0.60). The highest correlation between unobserved heterogeneity of low SES is between the heterogeneity in terms of completion and enrolling without working alongside (-0.84).

For high SES students, unobserved heterogeneity in terms of the decision to work during studies also has the highest variance (0.17) after the work decision but is substantially lower than for low SES students. In terms of correlated unobserved heterogeneity again completion and enrolling without working alongside exhibits the strongest correlation. However, for high SES students this correlation is positive (0.97).

Table C.1: Variance-covariance matrix of unobserved heterogeneity

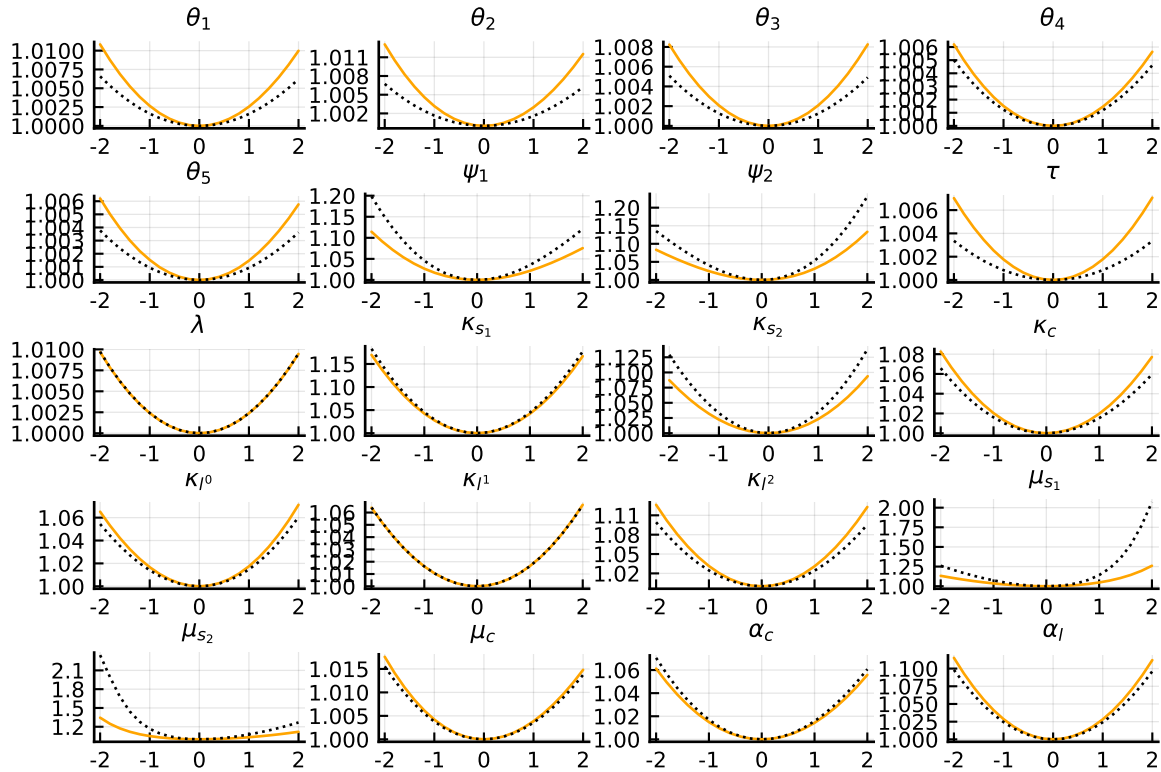
	Enroll & not work	Enroll & work	Complete	Work
<i>Low SES</i>				
Enroll and not work	0.602			
Enroll and work	0.023	0.061		
Complete	0.057	-0.031	0.023	
Work	0.156	0.016	-0.083	1.0
<i>High SES</i>				
Enroll and not work	0.167			
Enroll and work	0.005	0.092		
Complete	-0.143	0.114	0.15	
Work	-0.072	-0.063	-0.037	1.0



### C.3 Sensitivity Analysis

In Figure C.1 we look at how the criterion function behaves in the neighborhood of the optimum. In order to do so we incrementally decrease and increase parameters moving them up to two standard deviations away from their optimal value. While this is no definite proof that we have identified the global maximum, the smooth behavior around the identified optimum increases our confidence that indeed we have chosen optimal values for each parameter for both the low (orange solid line) and high (black dotted line) SES model.

Figure C.1: Response of criterion function to +/- 2 standard deviation change in each parameter

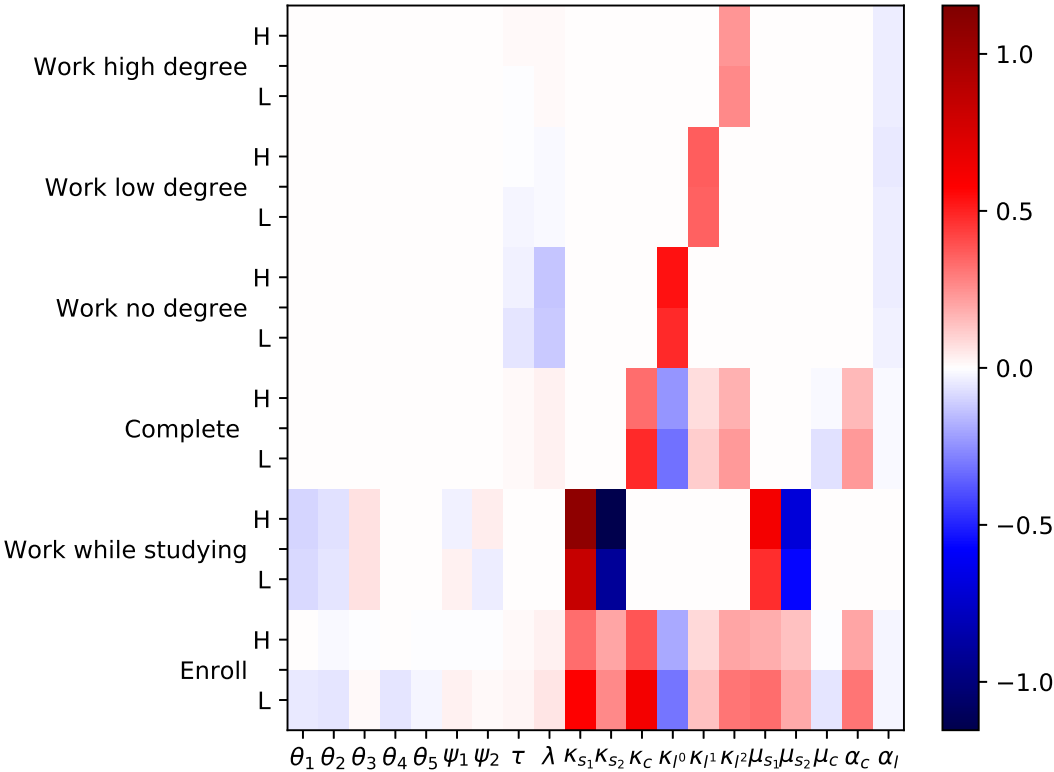


Notes: The y-axis shows the relative total error of the criterion function in response to changes to the parameters specified in the subtitles. The x-axis represents changes in terms of standard deviations. The orange solid line represents changes for low SES while the black dotted lines represents changes for high SES.

In Figure C.2 we see the relative change in each decision (y-axis) for low and high SES students as a response to a 0.5 standard-deviation increase in a parameter (x-axis).

Red cells indicate increases, whereas blue cells indicate a decrease. The darker a cell, the stronger a response, where the magnitude of the change in percentage terms can be gauged from the scale on the left hand side of the figure. The idea is to understand the sensitivity of key moments to changes in parameters.

Figure C.2: Response of decisions to 0.5 standard deviation increase in each parameter



Notes: Each cell represents the relative response of the model moment on the y-axis for low SES (bottom) and high SES (top) to a 0.5-standard deviation shift in the model parameter on the x-axis. The magnitude of the response is indicated by the shade of the cell as outlined in the scale on the right.

The first observation is that neither of the targets is too sensitive to any single parameter. We see that for both high and low SES students the maximum relative change is to the enrollment probability and the probability of working alongside studies when the respective direct utility intercepts  $\kappa_{s_2}$  and  $\kappa_{s_1}$  are shifted by 0.5 standard deviations. However, the relative change in the respective probabilities is around 1% and therefore not too volatile. The second observation is that we can see the difference

in the responsiveness of low and high SES students to changes in parameters, which are related to differences in perceived returns. For instance, increasing  $\theta_4$ , the parameter governing the importance of parental support in the choice, reduces attendance more for low SES students than for high SES students. Interestingly, it reduces attendance for both groups despite its positive sign. Remember that inputs are standardized, i.e. many students will have negative values for parental support in their choice, so a higher parameter value can lead to a lower utility from enrollment. The third observation is that the dynamic component of the model becomes apparent. If we increase  $\kappa_c$ , the intercept for the additional utility from completion, more students not only complete but also attend university in the first place. Now that more students are anticipating to complete university (because the utility from completion has increased), more students decide to enroll in university in the first place. Similarly, an increase in the value placed on job enjoyment  $\lambda$  increases the probability of working with a high degree. As a result high-degree jobs become more attractive, students become more likely to complete university and they are more likely to enroll in university in the first place.