

Optimal Financial Aid Policies for Students*

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Abstract

We study the optimal design of financial aid policies for college students. The total social benefits of college education exceed the private benefits because the government receives a share of the monetary returns in the form of income taxes. We underscore the policy implications of this fiscal externality in an optimal dynamic public finance framework. The model incorporates multidimensional heterogeneity, idiosyncratic risk and borrowing constraints. It matches key empirical results on college enrollment patterns, returns to education and enrollment elasticities. We find that a marginal increase in college subsidies in the US is at least 70 percent self-financing through the net-present value increase in future tax revenue. When targeting this increase to children in the lowest parental income tercile, it is more than self-financing with a fiscal return of 165 percent. The optimal Mirrleesian income tax schedule is barely affected, in particular if subsidies are set at their optimal level.

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1 Introduction

College graduates in the U.S. on average earn \$480,000 more over their lifecycle than individuals without a college degree (Autor 2014). According to the OECD, they contribute \$205,003 more to the government budget (OECD 2014). These numbers are not directly informative for policy because they reflect returns and selection. Nevertheless they suggest that income taxation can lift the social returns to education substantially above the private returns: individual college education decisions impose a *fiscal externality*. In this paper, we theoretically and quantitatively assess the implications of this fiscal externality for the efficient design of college education subsidies and income taxes.

To this end, we develop a dynamic optimal taxation framework, where individuals decide on college entry early in their life. One novel contribution to the literature following Mirrlees (1971) is that our model is very close to the recent structural labor literature on college education: individuals differ in ability, parental income and preferences, and face idiosyncratic uncertainty and borrowing constraints.¹ We quantify the model for the U.S. with micro-data and it matches key empirical features such as college education elasticities, marginal and average returns, as well as college education patterns in terms of ability and parental income. Despite the empirical plausibility of the model in terms of heterogeneity, we are able to fully characterize optimal policies and policy reforms analytically. We use a variational approach (Goloso, Tsyvinski, and Werquin 2014) to study education subsidies that do (or do not) condition on parental income and ability as well as arbitrarily nonlinear tax schedules.²

We use two complementary approaches: (i) we quantify the fiscal revenue effects of reforming current education policies in the US and (ii) we simulate the optimal design of these policy instruments given a Utilitarian welfare function. We now describe the two approaches in more detail.

Our first contribution is to show that even in the absence of any positive human capital externalities, there is a substantial case for increasing college education subsidies: the increase pays for itself to a large extent through higher discounted tax revenue in the future. Our baseline number for the U.S. is 70-87%. Thus, increasing college subsidies from their current level by one dollar costs the government effectively only between 13 and 30 Cents. Our formal

¹An early contribution with multi-dimensional heterogeneity is Kleven, Kreiner, and Saez (2009). Krueger and Ludwig (2013, 2015) study optimal college subsidies and linear income taxes in a large scale macroeconomic OLG model. Heathcote, Storesletten, and Violante (2014) characterize optimal tax progressivity when individuals are heterogenous w.r.t learning ability and disutility of work. Stantcheva (2015a) derives education and tax policies in an OLG model with multi-dimensional heterogeneity. Blundell, Dias, Meghir, and Shaw (2015) study the effect of tax credits and income support programs in a dynamic structural model with human capital accumulation.

²Our approach differs from the recent dynamic Mirrlees literature where more complex, history-dependent policy instruments are studied. See, e.g., Stantcheva (2014, 2015b) for models with endogenous human capital. In line with our approach in terms of policy instruments Best and Kleven (2013) study optimal history-independent taxation with on the job learning. Pavoni and Violante (2007) and Pavoni, Setty, and Violante (2012) elaborate the consequences of human capital depreciation during unemployment for the efficient design of welfare programs.

analysis transparently highlights the trade-offs underlying these numbers. The extent to which an increase in subsidies is self-financing increases in the ratio of marginal students (those who are induced to enroll because of an increase in subsidies) over inframarginal students (those that would have enrolled also in the absence of the increase), the marginal returns to college and the progressivity of the tax system. Motivated by these theoretical considerations, we consider an increase in subsidies for a group where the share of inframarginal students is small: children with low parental income. If the increase in subsidies is targeted to the lowest parental income tercile, the government obtains between \$1.00 and \$1.65 additional discounted tax revenue for each dollar of subsidies – increasing subsidies for this group is a free lunch. This is mainly driven by the lower share of inframarginal students among those children. The lower bounds of the values take into account dropout and adverse effects on parental labor supply. Allowing for general equilibrium effects on the fiscal externality makes the results, perhaps surprisingly, stronger.

Our second main contribution is to explore optimal policies in the spirit of Mirrlees (1971). In particular, we extend the Mirrlees approach to the analysis of education policies by allowing the government to design financial aid policies which are allowed to depend on parental income or academic merit. Despite the large amount of heterogeneity in the model, we can provide an analytical treatment of optimal nonlinear (and potentially education-dependent) taxes and education subsidies. For the optimal tax schedule, we only find small effects. Optimal marginal tax rates for given college subsidies (at their current level) are at most 1.2 percentage points lower as compared to the case with exogenous college enrollment decisions. For top incomes, the difference is negligible. If subsidies are set optimally, the differences are negligible at all income levels.

To describe our approach more formally, we consider a model where individuals decide whether to work directly after high school or postpone labor market entry and obtain a college degree.³ In the latter case, individuals make a borrowing decision during college. In either case, they face idiosyncratic wage risk that materializes after entering the labor market. The distribution from which they draw their wage depends on their education decision. After individuals have learned their wage, they make a labor leisure decision. Labor supply is endogenous with respect to the income tax system. Ex-ante, individuals differ with respect to ability, psychic costs of going to college and parental income. Ability determines the wage distributions, and therefore the returns to college as well as the opportunity costs of foregone earnings. Parental income matters because parents provide transfers to their children.

To quantify the model, we use data from the National Longitudinal Survey of Youth 1979 and 1997 (henceforth, NLSY79 and NLSY97). Parental income and ability are directly taken from the data. For the latter, we use the *Armed Forces Qualification Test Score* (AFQT). We estimate how parental income and the AFQT-score determine the other variables of our model.

³In the baseline model we abstract from college dropout. We extend our analysis for college subsidies and consider dropout in the appendix.

The most important objects for this are (i) the wage distributions conditional on AFQT and the education decision, (ii) parental transfers conditional on parental income and the education decision, as well as (iii) college grant receipt as a function of ability and parental income. We define each individual that obtained at least a bachelor degree as a college graduate. Based on these estimations, some standard assumptions on preferences and a detailed modeling of college education finance and income tax system, we can calculate the value functions for college and high school for each individual according to our model. Finally, we estimate the distribution of psychic costs for college with maximum likelihood to fit college decisions in the data. In line with the empirical literature discussed below, we choose parental education and AFQT as determinants of psychic costs. Our quantitative model can replicate key patterns on how college education varies with parental income and AFQT. Further, our model yields returns to college that are in line with the empirical literature (Card 1999, Oreopoulos and Petronijevic 2013, Zimmermann 2014).

Related Literature We derive our main theoretical results using intuitive tax perturbation methods, which go back to Piketty (1997) and Saez (2001) in the static context and have recently been generalized to dynamic settings by Golosov, Tsyvinski, and Werquin (2014). Concerning the results for optimal income taxes, we show how our formula is an extended version of the well known Diamond (1998) formula. Since college enrollment is modeled as a binary choice, our formal approach is similar to other optimal tax papers with both, intensive and extensive margin. Saez (2002) and Jacquet, Lehmann, and Van der Linden (2013) consider the labor force participation margin, Kleven, Kreiner, and Saez (2009) consider labor force participation of the secondary earner, Scheuer (2014) considers the occupational choice margin and Lehmann, Simula, and Trannoy (2014) consider migration.

Stantcheva (2014, 2015b), Kapicka and Neira (2015) and Findeisen and Sachs (2015a) consider models of optimal taxation with education in the “New Dynamic Public Finance” (NPDF) tradition with endogenous policy instruments. Our paper differs in that we explicitly consider college as an education decision and also analyze simpler policy instruments as opposed to solely focusing on the second-best. Further, our model is richer in terms of heterogeneity and very close the structural literature. Bovenberg and Jacobs (2005) is an important early paper in the literature. They show that education subsidies and redistributive taxes are “siamese twins” in that the former can counteract the distortion on the education margin of the latter. Bohacek and Kapicka (2008) extend the findings of Bovenberg and Jacobs (2005) to a dynamic setting. Richter (2009) studies a Ramsey setting and finds that qualified labor should be distorted less heavily than non-qualified labor. Gelber and Weinzierl (2012) show how the optimal tax system changes, when children’s abilities depend on parents’ financial resources. Studying age dependency as well as standard income taxes, Best and Kleven (2013) augment the canonical

optimal tax framework by incorporating career effects into a deterministic model. Weinzierl (2011) studies the welfare gains of age-dependent labor income taxation.

There is empirical evidence that financial variables like parental transfers and the monetary returns to college are not the only determinants of college entry. Different non-monetary variables like parental education and ability are typically taken into account for by a one-dimensional aggregate labelled psychic costs, see, e.g., Heckman, Lochner, and Todd (2006), Cunha, Karahan, and Soares (2011) and Johnson (2013). We model and quantify psychic costs in line with these papers.

One reason for the strong link between parental income and college enrollment are parental transfers. For various reasons, parents might subsidize the college decision of their children and the extent of such parental subsidies naturally varies with income. We follow Keane and Wolpin (2001) and Johnson (2013) and model parental transfers in a reduced form fashion. Finally, Johnson (2013) and Abbott, Gallipoli, Meghir, and Violante (2013) find that need based college grants are more cost-effective in increasing enrollment than merit-based grants because among low income individuals, the share of inframarginal students is low. This is related to our result that an increase in need-based financial aid is likely to be a free lunch.

2 Framework

2.1 Individual Heterogeneity, Preferences and Choices

Short Verbal Summary of the Model. Individuals start life as high school graduates and decide whether to obtain a college degree. They differ in innate ability, their preferences for college and parental income. If an individual decides against a college degree, she directly enters the labor market, draws a lifetime wage from a distribution that depends on her innate ability, and – conditional on the drawn wage – makes a labor-leisure decision. If going to college the individual has to make a savings decision. Once the individual graduates from college, she enters the labor market, draws a lifetime wage from a different distribution that also depends on her innate ability, and makes a labor-leisure decision.

Whether an individual attends college depends on innate ability through the returns to college and the opportunity costs of foregone earnings. It also depends on parental income, as we allow *parental transfers* to be determined by parental income. First, parental transfers matter because of (potentially binding) borrowing constraints. Second, parental transfers act as a price subsidy because parents make transfers contingent on the educational decision. Lastly, obtaining a college degree has a direct effect on utility.

Heterogeneity. The labels for the innate ability are $\theta \in \Theta := [\underline{\theta}, \bar{\theta}]$ and $\mathcal{I} \in \mathbb{R}_+$ for parental income. Additionally, we allow for heterogeneity in other characteristics X , where X is a N -dimensional vector. The joint probability density function (*pdf*) is given by $j(\theta, \mathcal{I}, X)$. Whereas

θ and $\mathcal{I}$ are variables that influence an individual's utility through the budget constraint, we assume that X contains variables that influence an individual's decision to go to college by factors outside the budget constraint. All these variables translate into one scalar which we denote *psychic costs* of going to college. Psychic costs κ are thus a function of X . We allow also θ to influence these psychic costs. Hence, we can write $\kappa = z(\theta, X)$.⁴ The individual college attendance decision then depends on the triple $(\theta, \mathcal{I}, \kappa)$. The joint density function $m(\theta, \mathcal{I}, \kappa)$ is given by

$$m(\theta, \mathcal{I}, \kappa) = \int_{\{x \in X | k(\theta, x) = \kappa\}} j(\theta, \mathcal{I}, x) dx.$$

We denote the marginal density of ability as $f(\theta)$ (the associated *cdf* is $F(\theta)$), the density of parental income conditional on innate ability as $k(\mathcal{I}|\theta)$ (the associated *cdf* is $K(\mathcal{I}|\theta)$) and the density of psychic costs conditional on ability and parental income as $h(\kappa|\theta, \mathcal{I})$ (the associated *cdf* is $H(\kappa|\theta, \mathcal{I})$). They are related to the joint density $m(\theta, \mathcal{I}, \kappa)$ by

$$m(\theta, \mathcal{I}, \kappa) = f(\theta)k(\mathcal{I}|\theta)h(\kappa|\theta, \mathcal{I}).$$

For notational simplicity, we often denote the conditional densities as $h_{\theta, \mathcal{I}}(\kappa)$ and $k_{\theta}(\mathcal{I})$ and the conditional distribution functions as $H_{\theta, \mathcal{I}}(\kappa)$ and $K_{\theta}(\mathcal{I})$.

After individuals make the decision to go to college or not, their wage $\omega \in \Omega := [\underline{\omega}, \bar{\omega}]$ materializes. If they go to college – i.e. if they are college graduates – they draw their wage from a distribution with *cdf* $G_{\theta}^{co}(\omega)$ and density $g_{\theta}^{co}(\omega)$. Otherwise – if they are high-school graduates – they draw their wage from a distribution with *cdf* $G_{\theta}^{hs}(\omega)$ and density $g_{\theta}^{hs}(\omega)$. Wage distributions are only influenced by ability and education. In the theoretical part of this paper, we do not make any further assumptions on these wage distributions; note that they encompass labor market returns from innate ability, returns to college, a possible complementarity between ability and schooling as well as the degree of uncertainty.

Preferences. We assume quasi-linear preferences of the form $U(c - v(\frac{y}{\omega}))$, where c denotes consumption, y denotes gross income, $U(\cdot)$ is a concave transformation and $v(\cdot)$, disutility of labor effort, is a strictly convex transformation that satisfies $v(0) = 0$. We make the quasi-linearity assumption for tractability.⁵ This assumption allows us to solve for optimal nonlinear income taxes à la Mirrlees (1971) despite the large amount of heterogeneity. Concretely, the labor supply decision of an individual only depends on ω and policies; it is independent of the amount of non-labor income or wealth. This implies that the decision is independent of

⁴Our modeling of psychic costs is comparable and close to Cunha, Heckman, and Navarro (2005), Heckman, Lochner, and Todd (2006), Cunha, Karahan, and Soares (2011), Navarro (2011) and Abbott, Gallipoli, Meghir, and Violante (2013), among others.

⁵The empirical literature using micro data sets has typically found only small income effects on labor supply (see Gruber and Saez (2002) for the US or a recent paper by Kleven and Schultz (2012) using the universe of Danish tax records). In macroeconomics, this class of preferences has shown to be very useful in matching business cycle moments (Greenwood, Hercowitz, and Huffman 1988, Mendoza and Yue 2012).

X , θ and $\mathcal{I}$.⁶ This renders the screening problem one-dimensional. Further, we assume that individuals are expected utility maximizers. Denote by $c_{co}(\omega, \theta, \mathcal{I})$ and $y_{co}(\omega)$ consumption and before tax earnings of a college graduate with wage ω , innate ability θ and parental income $\mathcal{I}$, and by $c_e(\theta, \mathcal{I})$ the respective consumption during college.⁷ Expected utility of a college graduate with innate ability θ and parental income $\mathcal{I}$ is given by

$$\beta^{co1}U(c_e(\theta, \mathcal{I})) - \kappa + \beta^{co2} \int_{\Omega} U \left(c_{co}(\omega, \theta, \mathcal{I}) - v \left(\frac{y_{co}(\omega)}{\omega} \right) \right) dG_{\theta}^{co}(\omega),$$

where β^{co1} and β^{co2} capture discounting and the different length of periods. Assume that life after the college entry decision lasts T periods, college takes T_e periods and individuals' yearly discount factor is β . Then we can think of $\beta^{co1} = \sum_{t=1}^{T_e} \beta^{t-1}$ and $\beta^{co2} = \sum_{t=T_e+1}^T \beta^{t-1}$. As is a standard assumption in the literature on college enrollment, psychic costs κ and consumption enter utility separably. Note that κ can also take negative values, which then reflects utility benefits from attending college. κ should be interpreted as a one-dimensional aggregate of different factors (i.e. the vector X) that determine the college entry decision by factors outside the budget constraint. In the empirical section of this paper, we will be more explicit about these factors.

Denote by $c_{hs}(\omega, \mathcal{I})$ consumption of a high school graduate with wage ω and parental income $\mathcal{I}$. Note that consumption is not a function of θ in this case, because high school graduates do not make a savings decision – their life is static. $y_{hs}(\omega)$ are before tax earnings. Expected utility of a high school graduate with innate ability θ and parental income $\mathcal{I}$ is then given by

$$\beta^{hs} \int_{\Omega} U \left(c_{hs}(\omega, \mathcal{I}) - v \left(\frac{y_{hs}(\omega)}{\omega} \right) \right) dG_{\theta}^{hs}(\omega),$$

where $\beta^{hs} = \beta^{co1} + \beta^{co2}$ captures the length of the labor market period of high school graduates. Given that life of a high school graduate captures no dynamics, it might seem unnecessary to introduce β^{hs} . However, for the decision of going to college or not, it is important to capture the time aspect to address foregone earnings as opportunity costs of college.

Choices. Individuals first decide on whether to go to college. If individuals attend college, they also make a savings decision. Besides the psychic costs (or benefits) associated with going to college and the foregone earnings, individuals also have to pay tuition fees C . Once they have entered the labor market, individuals make a labor-leisure decision conditional on their realized wage. We do not model college dropout, but show how our results extend to this case in Appendix A.2.2. While parents are not explicitly modeled, we capture family transfers

⁶In Findeisen and Sachs (2015b), we elaborate this idea to study simple income taxation in a dynamic stochastic environment, i.e. we characterize income taxes that are simple in the sense that they are based on current annual income as opposed to the whole history of incomes.

⁷Consumption decisions are independent of κ because psychic costs enter the utility function in a separable way. As explained above, earnings decisions are independent of θ and $\mathcal{I}$ because we abstract from income effects.

in a reduced-form manner. Transfers vary with parental income and also with the education decision. We denote the transfers functions by $tr_{hs}(\mathcal{I})$ and $tr_{co}(\mathcal{I})$. Next, we formally state the individuals' problems and the associated value functions for given income taxes and education subsidies.

2.2 Equilibrium Given Policies

We next introduce taxes and education subsidies into the model. The tax functions that we consider are potentially education dependent. We denote by $\mathcal{T}^{hs}(y)$ the tax schedule for high school graduates and by $\mathcal{T}^{co}(y)$ the tax schedule for college graduates. We analyze optimal education independent taxes and can just write $\mathcal{T}(y)$. $\mathcal{G}(\theta, \mathcal{I})$ is the subsidy or grant that college students receive; the dependence on θ and $\mathcal{I}$ reflects merit and need-based financial aid.

For a type $(\theta, \mathcal{I})$, the value function of being a high school graduate is given by

$$V_{hs}(\theta, \mathcal{I}) = \max_{y^{hs}(w)} \int_{\underline{w}}^{\bar{w}} \beta^{hs} U \left(y^{hs}(w) + \frac{tr_{hs}(\mathcal{I})}{\sum_{t=1}^T \frac{1}{(1+r)^{t-1}}} - \mathcal{T}_{hs}(y^{hs}(w)) - v \left(\frac{y^{hs}(w)}{w} \right) \right) dG_{\theta}^{hs}(w). \quad (1)$$

A high school graduate essentially solves a static problem as there is no more risk after the initial wage draw. The period length has to be taken into account when transforming the parental transfer into consumption. In the following, we assume $\beta(1+r) = 1$, which implies $\frac{tr_{hs}(\mathcal{I})}{\sum_{t=1}^T \frac{1}{(1+r)^{t-1}}} = \frac{tr_{hs}(\mathcal{I})}{\beta^{hs}}$. For the college graduate, the problem is slightly more involved:

$$V_{co}(\theta, \mathcal{I}) = \max_{c_e, c^{co}(w), y^{co}(w)} \beta^{co1} U(c_e) + \beta^{co2} \int_w U \left(c^{co}(w) - v \left(\frac{y^{co}(w)}{w} \right) \right) dG_{\theta}^{co}(w) \quad (2)$$

subject to

$$\forall w : \beta^{co2} c^{co}(w) \leq \beta^{co2} [y^{co}(w) - \mathcal{T}^{co}(y^{co}(w))] + \beta^{co1} s$$

and

$$c_e = \frac{tr_{co}(\mathcal{I})}{\beta^{co1}} + \mathcal{G}(\theta, \mathcal{I}) - C - s,$$

and

$$s \geq \bar{s}.$$

$-s$ is the amount individuals borrow during education. The last constraint implies that individuals are borrowing constrained.⁸

⁸One could easily generalize the approach and assume that the borrowing limit depends on θ or $\mathcal{I}$. Given that we consider $\bar{s}$ being independent of these variables in our quantitative implementation of the model, we stick to this case already at this point of the analysis.

We now turn to the college entry decision. An individual of type $(\theta, \mathcal{I}, \kappa)$ goes to college whenever

$$V_{co}(\theta, \mathcal{I}) - \kappa \geq V_{hs}(\theta, \mathcal{I}).$$

From this expression it follows that there exists a unique threshold of κ for each ability level that divides individuals into college and high school graduates. This threshold is defined by

$$\tilde{\kappa}(\theta, \mathcal{I}) = V_{co}(\theta, \mathcal{I}) - V_{hs}(\theta, \mathcal{I}) \quad (3)$$

for each $(\theta, \mathcal{I})$.

To ease notation, we define a set of cross sectional wage distributions. First, we define cross sectional distributions conditional on education:

$$G^{co}(\omega) = \int_{\Theta} \int_{\mathbb{R}_+} G_{\theta}^{co}(\omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta) \quad (4)$$

and

$$G^{hs}(\omega) = \int_{\Theta} \int_{\mathbb{R}_+} G_{\theta}^{hs}(\omega) (1 - H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I}))) dK_{\theta}(\mathcal{I}) dF(\theta).$$

Note that G^{co} and G^{hs} are not classical *cumulative distribution functions* because $G^{co}(\bar{\omega}) < 1$ and $G^{hs}(\bar{\omega}) < 1$. Finally, the overall wage distribution is given by

$$G(\omega) = G^{co}(\omega) + G^{hs}(\omega).$$

G is a classical *cdf* because we have $G(\bar{\omega}) = 1$. Note that these wage distributions are endogenous with respect to the tax system. For ease of notation we suppress dependence of these functions with respect to policies.

2.3 Responses Along the College Margin to Policy Changes

In this subsection, we elaborate how the individual college decision responds to changes in policies. Modeling – and quantitatively assessing – behavioral responses along the college margin and assessing the policy implications is the central contribution of this paper. Individuals can respond to policies along three margins: the labor supply margin, the savings margin and the education margin. As individual preferences satisfy quasi-linearity, labor supply will not respond to changes in the absolute value of taxes (and therefore not to changes in the lump sum elements of the tax schedules), but only to changes in marginal tax rates. These responses and their implications for policy design are standard considerations in public economics and well understood.⁹ Therefore, we refrain from analyzing labor supply responses in detail. We

⁹See Piketty and Saez (2013) for a survey.

also do not discuss the responses of savings with respect to policies because we do not consider loan subsidies and related policy instruments.

2.3.1 Definition of Graduation Semi-Elasticities

College enrollment is responsive to changes in subsidies and the absolute value of taxes. We define *enrollment semi-elasticities* with respect to subsidies and taxes and start with the former:

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$$\xi^{\mathcal{G}}(\theta, \mathcal{I}) \equiv \frac{\frac{\partial \tilde{\kappa}(\theta, \mathcal{I})}{\partial \mathcal{G}} h_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) k(\mathcal{I}|\theta) f(\theta)}{H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) k(\mathcal{I}|\theta) f(\theta)} = \frac{\beta^{co1} U'(c_e(\theta, \mathcal{I})) h_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I}))}{H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I}))} > 0.$$

$\xi^{\mathcal{G}}(\theta, \mathcal{I})$ says by what percentage the share of students of type $(\theta, \mathcal{I})$ increases due to a \$1 increase in grants. The numerator captures the mass of marginal individuals that are induced to go to college due to the subsidy increase. To obtain a semi-elasticity, the mass of those marginal individuals of type $(\theta, \mathcal{I})$ is divided by the share of inframarginal students of type $(\theta, \mathcal{I})$. The equality follows from the fact that $\frac{\partial \tilde{\kappa}(\theta, \mathcal{I})}{\partial \mathcal{G}} = \beta^{co1} U'(c_e(\theta, \mathcal{I}))$ which one obtains from differentiation of (3). We define semi-elasticities in such a way that a positive semi-elasticity implies an increase in the college graduation rate. Finally note that semi-elasticities tend to be higher if individuals are borrowing constrained. Intuitively, consumption during college is lower in the presence of credit constraints, which increases the value of $U'(c_e(\theta, \mathcal{I}))$.

3 Optimal Financial Aid Policies: Theory

In this section, we characterize the optimal design of financial student aid. It will be useful to first work through a partial reform approach, looking at the effects of a small change in policies. The tax reform methodology has been used in Public Finance to study optimal taxation by Piketty (1997) and Saez (2001) in the static context and has recently been generalized to dynamic settings by Golosov, Tsyvinski, and Werquin (2014). We will now employ a similar perturbation approach to study optimal financial aid policies.

3.1 Fiscal Effects

We first derive the fiscal effect of marginally increasing all grants by $d\mathcal{G}$. This lays the ground for the full optimum which we consider in the next section. First of all, a grant increase has a *mechanical* effect on revenue because additional money is paid to all students. This impact on the government budget is given by

$$M^{\mathcal{G}} = -d\mathcal{G} \times \beta^{co1} G^{co}(\bar{\omega}), \quad (5)$$

recall that $G^{co}(\bar{\omega})$ (as defined in (4)) is the share of college graduates. Everything is multiplied by β^{co1} because we define the grant increase such that it is equally increased in all years during college.

Apart from this mechanical revenue effect, the small grant increase also has a behavioral effect on public funds. Higher grants induce some students to go to college that would not have gone otherwise – we label these students as marginal students. These marginal students matter for the government budget since subsidies have to be paid to them and since they will pay a different amount of taxes over their life cycle. This change depends on innate ability and parental income because (i) grants depend on ability as well as parental income and (ii) additional tax revenue due to returns depends on ability:

$$\Delta\mathcal{T}(\theta, \mathcal{I}) = \beta^{co2} \int_{\Omega} \mathcal{T}^{co}(y_{co}(\omega)) dG_{\theta}^{co}(\omega) - \beta^{co1} \mathcal{G}(\theta, \mathcal{I}) - \beta^{hs} \int_{\Omega} \mathcal{T}^{hs}(y_{hs}(\omega)) dG_{\theta}^{hs}(\omega). \quad (6)$$

The impact on the government budget of the increase in college graduation – that we call the college graduation effect – is given by:

$$CG^{\mathcal{G}} = d\mathcal{G} \times \int_{\Theta} \int_{\mathbb{R}_+} \Delta\mathcal{T}(\theta, \mathcal{I}) \xi^{\mathcal{G}}(\theta, \mathcal{I}) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta), \quad (7)$$

where $\xi^{\mathcal{G}}(\theta, \mathcal{I})$ is the semi-elasticity of college graduation with respect to grants as defined in Section 2.3.1. The overall impact on public funds of increasing $\mathcal{G}(\theta, \mathcal{I})$ is then given by:

$$\Gamma^{\mathcal{G}} = CG^{\mathcal{G}} + M^{\mathcal{G}}. \quad (8)$$

If $\Gamma^{\mathcal{G}}$ is negative, this implies that increasing public subsidies is not a free lunch. Otherwise, increasing subsidies is (dynamically) self-financing as the additional tax revenue outweighs the direct fiscal costs of the increase in subsidies. Dividing by the negative of the mechanical effect, one gets

$$\gamma^{\mathcal{G}} = \frac{\Gamma^{\mathcal{G}}}{\beta^{co1} d\mathcal{G} G^{co}(\bar{\omega})}. \quad (9)$$

This expression can be interpreted as the rate of return on one dollar invested in additional college subsidies. If it takes the value .2, it says that the government gets \$1.20 in additional tax revenue for one marginal dollar invested into college subsidies. If it is -.5, it implies that the government gets 50 Cents back for each dollar invested. Inserting (7) and (8) into (9) gives

$$\gamma^{\mathcal{G}} = \frac{\int_{\theta} \int_{\mathcal{I}} \Delta e(\theta, \mathcal{I}) \Delta\mathcal{T}(\theta, \mathcal{I}) d\mathcal{I} d\theta}{\beta^{co1} G^{co}(\bar{\omega})} - 1, \quad (10)$$

where $\Delta e(\theta, \mathcal{I}) \equiv \xi^{\mathcal{G}}(\theta, \mathcal{I}) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) k_{\theta}(\mathcal{I}) f(\theta)$ is the mass of marginal students of type $(\theta, \mathcal{I})$.

Next, define the normalized overall share of marginal students $\bar{\Delta}e = \int_{\theta} \int_{\mathcal{I}} \Delta e(\theta, \mathcal{I}) d\mathcal{I} d\theta / \beta^{co1}$. It is normalized such that it refers to a marginal increase of $1/\beta^{co1}$ of grants, i.e. a grant increase that costs the government exactly one dollar for each inframarginal student. Further, we define the average fiscal externality for marginal students $\bar{\Delta}\mathcal{T} = \frac{\int_{\theta} \int_{\mathcal{I}} \Delta e(\theta, \mathcal{I}) \Delta \mathcal{T}(\theta, \mathcal{I}) d\mathcal{I} d\theta}{\int_{\theta} \int_{\mathcal{I}} \Delta e(\theta, \mathcal{I}) d\mathcal{I} d\theta}$. This simplifies notation to get

$$\gamma^{\mathcal{G}} = \frac{\bar{\Delta}e \times \bar{\Delta}\mathcal{T}}{G^{co}(\bar{\omega})} - 1.$$

Thus, whether the return on one dollar of additional college subsidies is positive depends on three measures: (i) the share of inframarginal students $G^{co}(\bar{\omega})$, (ii) the share of marginal students $\bar{\Delta}e$ and (iii) the average fiscal externality per marginal student $\bar{\Delta}\mathcal{T}$. This formula does not depend on the exact modeling specification but can instead be considered as a *sufficient-statistics* formula (Chetty 2009).

We also assess the profitability of public educational investment for students with different parental income (need-based aid) and innate ability (merit based grants). For example, if one thinks about increasing grants only for individuals with parental income below $\mathcal{I}^*$, formula (10) becomes

$$\gamma_{\mathcal{I}^*}^{\mathcal{G}} = \frac{\bar{\Delta}e_{\mathcal{I}^*} \times \bar{\Delta}\mathcal{T}_{\mathcal{I}^*}}{G_{\mathcal{I}^*}^{co}(\bar{\omega})} - 1, \quad (11)$$

where all variables are adjusted to take into account that the reform only affects a subset of the population.¹⁰ Crucially, the behavioral responses, marginal returns and student shares may differ a lot across parental income groups.

3.2 Optimal Financial Aid Policies

The objective of the government is:

$$\begin{aligned} & \int_{\Theta} \int_{\mathbb{R}_+} \int_{\kappa}^{\tilde{\kappa}(\theta, \mathcal{I})} (V_{co}(\theta, \mathcal{I}) - \kappa) d\tilde{H}_{\theta, \mathcal{I}}(\kappa) d\tilde{K}_{\theta}(\mathcal{I}) d\tilde{F}(\theta) \\ & + \int_{\Theta} \int_{\mathbb{R}_+} \int_{\tilde{\kappa}(\theta, \mathcal{I})}^{\bar{\kappa}} V_{hs}(\theta, \mathcal{I}) d\tilde{H}_{\theta, \mathcal{I}}(\kappa) d\tilde{K}_{\theta}(\mathcal{I}) d\tilde{F}(\theta), \end{aligned}$$

¹⁰Formally, the terms are defined s.t.

$$\gamma_{\mathcal{I}^*}^{\mathcal{G}} = \frac{\int_{\theta} \int_0^{\mathcal{I}^*} \Delta e(\theta, \mathcal{I}) \Delta \mathcal{T}(\theta, \mathcal{I}) d\mathcal{I} d\theta}{\int_{\theta} \int_0^{\mathcal{I}^*} \int_{\Omega} dG_{\theta}(\omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta)} - 1.$$

where $\tilde{\kappa}(\theta, \mathcal{I})$ is defined by (3) and $V_{co}(\theta, \mathcal{I})$ and $V_{hs}(\theta, \mathcal{I})$ are defined by (2) and (1). $\tilde{H}_{\theta, \mathcal{I}}(\kappa)$, $\tilde{K}_{\theta}(\mathcal{I})$ and $\tilde{F}(\theta)$ denote the cumulated Pareto weights. If, e.g., $\tilde{H}_{\theta, \mathcal{I}}(\kappa) = H_{\theta, \mathcal{I}}(\kappa)$, $\tilde{K}_{\theta}(\chi) = K_{\theta}(\chi) \forall \theta, \chi$ and $\tilde{F}(\theta) = F(\theta) \forall \theta$; the social planner is Utilitarian.

Policies have to satisfy a government budget constraint:

$$\begin{aligned} & \int_{\Theta} \int_{\mathbb{R}_+} \left(-\beta^{co1} \mathcal{G} + \beta^{co2} \int_{\Omega} \mathcal{T}^{co}(y_{co}(\omega)) dG_{\theta}^{co}(\omega) \right) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta) \\ & + \int_{\Theta} \int_{\mathbb{R}_+} \beta^{hs} \int_{\Omega} \mathcal{T}^{hs}(y_{hs}(\omega)) dG_{\theta}^{hs}(\omega) (1 - H_{\theta, \mathcal{I}}(\tilde{\kappa})) dK_{\theta}(\mathcal{I}) dF(\theta) \geq 0. \end{aligned} \quad (12)$$

We label ρ the multiplier on constraint (12). It will reflect the marginal value of public funds.

We make use of welfare weights that are endogenous with respect to policies. We define the welfare weight of an individual of type $(\theta, \mathcal{I}, \kappa, \omega)$ with $\kappa < \tilde{\kappa}(\theta, \mathcal{I})$ as¹¹

$$\mathcal{W}(\theta, \kappa, \mathcal{I}, \omega) = \frac{\beta^{co}}{\rho} U' \left(c_{co}(\theta, \mathcal{I}, \omega) - v \left(\frac{y_{co}(\omega)}{\omega} \right) \right) \frac{\tilde{h}_{\theta, \mathcal{I}}(\kappa) \tilde{k}_{\theta}(\mathcal{I}) \tilde{f}(\theta)}{\tilde{h}_{\theta, \mathcal{I}}(\tilde{\kappa}) \tilde{k}_{\theta}(\mathcal{I}) \tilde{f}(\theta)}.$$

The welfare weights are endogenous because they depend on the marginal utility of consumption and therefore on policies. Marginally increasing consumption for individuals of type $(\theta, \mathcal{I}, \kappa, \omega)$ increases welfare in units of public funds by $\mathcal{W}(\theta, \mathcal{I}, \kappa, \omega) \times g_{\theta}^{co}(\omega) h_{\theta, \mathcal{I}}(\kappa) k_{\theta}(\mathcal{I}) f(\theta)$. Or, equivalently, increasing consumption of $(\theta, \mathcal{I}, \kappa, \omega)$ by one dollar is as desirable as obtaining $\mathcal{W}(\theta, \mathcal{I}, \kappa, \omega) \times g_{\theta}^{co}(\omega) h_{\theta, \mathcal{I}}(\kappa) k_{\theta}(\mathcal{I}) f(\theta)$ additional dollars of public funds. We now also define two aggregated welfare weights. The welfare weights of all individuals with wage ω and with/without a college degree are given by:

$$\mathcal{W}^{co}(\omega) = \frac{1}{\rho} \frac{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\tilde{\kappa}(\theta, \mathcal{I})}^{\tilde{\kappa}(\theta, \mathcal{I})} U' \left(c_{co}(\theta, \mathcal{I}, \omega) - v \left(\frac{y_{co}(\omega)}{\omega} \right) \right) g_{\theta}^{co}(\omega) d\tilde{H}_{\theta, \mathcal{I}}(\kappa) d\tilde{K}_{\theta}(\mathcal{I}) d\tilde{F}(\theta)}{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\tilde{\kappa}(\theta, \mathcal{I})}^{\tilde{\kappa}(\theta, \mathcal{I})} g_{\theta}^{co}(\omega) dH_{\theta, \mathcal{I}}(\kappa) dK_{\theta}(\mathcal{I}) dF(\theta)} \quad (13)$$

and

$$\mathcal{W}^{hs}(\omega) = \frac{1}{\rho} \frac{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\tilde{\kappa}(\theta, \mathcal{I})}^{\tilde{\kappa}(\theta, \mathcal{I})} U' \left(c_{hs}(\mathcal{I}, \omega) - v \left(\frac{y_{hs}(\omega)}{\omega} \right) \right) g_{\theta}^{hs}(\omega) d\tilde{H}_{\theta, \mathcal{I}}(\kappa) d\tilde{K}_{\theta}(\mathcal{I}) d\tilde{F}(\theta)}{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\tilde{\kappa}(\theta, \mathcal{I})}^{\tilde{\kappa}(\theta, \mathcal{I})} g_{\theta}^{hs}(\omega) dH_{\theta, \mathcal{I}}(\kappa) dK_{\theta}(\mathcal{I}) dF(\theta)} \quad (14)$$

Based on that, the social marginal welfare weight of all individuals with wage ω is given by $\mathcal{W}(\omega) = \mathcal{W}^{co}(\omega) + \mathcal{W}^{hs}(\omega)$.

The above weights were defined for the working population. We also define the welfare weight of a college student of type $(\theta, \mathcal{I}, \kappa)$ by

¹¹For $\kappa > \tilde{\kappa}(\theta, \mathcal{I})$, we would have to replace subscript *co* by *hs*.

$$\mathcal{W}^{st}(\theta, \mathcal{I}, \kappa) = \frac{1}{\rho} U'(c_e(\theta, \mathcal{I})) \frac{\tilde{h}_{\theta, \mathcal{I}}(\kappa) \tilde{k}_{\theta}(\mathcal{I}) \tilde{f}(\theta)}{h_{\theta, \mathcal{I}}(\kappa) k_{\theta}(\mathcal{I}) f(\theta)}.$$

Similar as the aggregated worker weights, we also define an aggregated student weight

$$\mathcal{W}^{st} = \frac{1}{\rho} \frac{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\underline{\kappa}}^{\tilde{\kappa}(\theta, \mathcal{I})} U'(c_e(\theta, \mathcal{I})) d\tilde{H}_{\theta, \mathcal{I}}(\kappa) d\tilde{K}_{\theta}(\mathcal{I}) d\tilde{F}(\theta)}{\int_{\Theta} \int_{\mathbb{R}_+} \int_{\underline{\kappa}}^{\tilde{\kappa}(\theta, \mathcal{I})} dH_{\theta, \mathcal{I}}(\kappa) dK_{\theta}(\mathcal{I}) dF(\theta)}. \quad (15)$$

3.2.1 Type Independent Policies

We start with the formula for the optimal college grant:

Proposition 1. *The optimal college grant $\mathcal{G}$ satisfies:*

$$\mathcal{G} = \frac{\int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \eta^{co}(\theta, \mathcal{I}) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta)}{\beta^{col} G^{co}(\bar{w}) (1 - \mathcal{W}^{st})}, \quad (16)$$

where $\eta^{co}(\theta, \mathcal{I}) = \xi^{\mathcal{G}}(\theta, \mathcal{I}) \cdot \mathcal{G}$ is the elasticity of college graduation w.r.t. the education subsidy for individuals of type $(\theta, \mathcal{I})$ and $\mathcal{W}^{st}$ is the average social welfare weight of a student as defined in (15).

The intuition is very similar to the intuition behind the revenue gains from increasing college subsidies. The numerator captures the *fiscal externality*, i.e. the gains from subsidizing college. The stronger individuals respond to subsidies (captured by the elasticity $\eta^{co}(\theta, \mathcal{I})$), the larger subsidies should be. In addition, the higher taxes are and the higher the returns to college are, the larger is $\Delta \mathcal{T}(\theta, \mathcal{I})$ and the larger subsidies should be.

The denominator reflects the cost of increasing college subsidies. *Ceteris paribus*, higher college graduation rates $G^{co}(\bar{w})$ are associated with a lower subsidy, as all graduates receive the subsidy. But this cost term is corrected by the factor $1 - \mathcal{W}^{st}$ since the planner also values the consumption of each marginal student by $\mathcal{W}^{st}$ in terms of public funds. Note also that this formula for the optimal education subsidy does not only apply for the optimal income tax but also for any given suboptimal income tax function.¹²

4 Estimation and Calibration

4.1 Data & Procedure

We use two data sets to bring our model to the data: the National Longitudinal Survey of Youth 79 and 97 (henceforth NLSY79 and the NLSY97). A big advantage of these data sets,

¹²Relatedly, Lawson (2014) also studies optimal college tuition subsidies to internalize fiscal externalities. Jacobs, Schindler, and Yang (2012) study optimal linear education subsidies in the presence of a fiscal externality and elaborate on the role of earnings risk.

Table 1: Quantification of the Model

Object	Description	Procedure/Target
$F(\theta)$	Distribution of innate abilities	AFQT scores from NSLY97
$K_\theta(\mathcal{I})$	Conditional distribution of parental income	Taken from NSLY97
ω	Individual wage	Calibration from gross income as in Saez (2001)
$G_\theta^{hs}(\omega)$	Conditional Wage Distribution High-school	Estimated from regressions
$G_\theta^{co}(\omega)$	Conditional Wage Distribution College	Estimated from regressions
$tr_{hs}(\mathcal{I})$	Conditional Transfer Distribution High-school	Estimated from regressions
$tr_{co}(\mathcal{I})$	Conditional Transfer Distribution College	Estimated from regressions
$H_{\theta,\mathcal{I}}(\kappa)$	Psychic costs distributions	Maximum Likelihood
Utility Function: $\frac{\left(C - \frac{l^{1+\varepsilon}}{1+\varepsilon}\right)^{1-\gamma}}{1-\gamma}$ $\varepsilon=0.5$ Labor Supply Elasticity $\gamma=2$ Risk aversion		
Current Policies		
$\bar{s}$	Stafford Loan Maximum	Value in year 2000
$\mathcal{T}(y)$	Current Tax Function	Gouveia-Strauss Approximation from Guner et al. (2013)
$\mathcal{G}(\theta, \mathcal{I})$	Need- and Merit Based Grants	Estimated from regressions

which has been exploited in many previous papers, is that they contain the *Armed Forced Qualification Test Score* (AFQT-score) for most individuals, which is a cognitive ability score for high school students that is conducted by the US army. The test score is a good signal for ability. Cunha, Karahan, and Soares (2011), e.g., show that it is the most precise signal for innate ability among comparable scores in other data sets.

The NLSY97 is the baseline for our analysis. To quantify the joint distribution of parental income and ability, we take the cross sectional joint distribution in our sample. We then estimate how these variable map into the other variables (parental transfers, wages, grants, psychic costs) of the model. Since individuals in the NLSY97 set are born between 1980 and 1984, not enough information about their earnings is available to quantify the conditional wage distributions. To obtain these conditional wage distributions, we therefore use the NLSY79 data as this data set contains more information about labor market outcomes – individuals are born between 1957 and 1964. Combining both data sets in such a way has proven to be a fruitful way in the literature to overcome the limitations of each individual data set, see Johnson (2013) and Abbott, Gallipoli, Meghir, and Violante (2013). The underlying assumption is that the relation between AFQT and wages has not changed over that time period. We use the method of Altonji, Bharadwaj, and Lange (2011) to make the AFQT-scores comparable between the two samples and different age groups.

Finally, we define an individual as a college graduate if she has completed at least a bachelor's degree. Otherwise she counts as a high school graduate. Since individuals in the NLSY97 turn

18 years old between 1998 and 2002, we express all US-dollar amounts in year 2000 dollars. To quantify our model, we proceed as follows:

1. We calibrate and preset a few parameters and policies, described in Section 4.2.
2. We estimate $G_{\theta}^{co}(\omega)$, $G_{\theta}^{hs}(\omega)$ in 4.3.
3. Transfer function $tr_{co}(\mathcal{I})$, $tr_{hs}(\mathcal{I})$ and grant receipt are estimated $\mathcal{G}(\theta, \mathcal{I})$ by OLS regressions. For brevity, details of our procedure for transfers and grants are relegated to the appendix. Economically, the most important results for parental transfers is the strong dependence on education choice by the child. This contingency of parental transfers acts as a price subsidy for college. On top, we recover the well-know positive correlation between parental income and transfers. For grants, we find a strong negative effects of parental transfers on financial aid receipt at the extensive and intensive margin. Additionally, we can capture merit based grants by the conditional correlation of AFQT scores with grant receipt.
4. Based on that, we calculate $V_{hs}(\theta, \mathcal{I})$ and $V_{co}(\theta, \mathcal{I})$ for each individual and estimate the distribution of psychic costs with maximum likelihood in Section 4.4.

4.2 Preset Parameters and Policies

We assume the following utility function

$$U = \frac{\left(C - \frac{l^{1+\varepsilon}}{1+\varepsilon}\right)^{1-\gamma}}{1-\gamma},$$

where we set $\varepsilon = 2$, which implies a compensated labor supply elasticity of .5, and $\gamma = 2$. Micro-evidence suggests that the compensated elasticity is probably lower, around .33 (Chetty, Guren, Manoli, and Weber 2011). Given that our elasticity reflects the labor supply responsiveness over the life cycle, we take a larger value of .5. We assume that college takes 4.5 years (i.e. $T_e = 4.5$) and assume that individuals spend 43.5 or 48 years on the labor market depending on whether they went to college before. The choice of 4.5 years for degree completion corresponds to the average years to graduation we observe in the NLSY97, which is 4.57 years. This lines up well with numbers from other sources, for example, from the National Center for Education Statistics (NCES).¹³ We set the risk free interest rate to 3%, i.e. $R = 1.03$ and assume that individuals' discount factor is $\beta = \frac{1}{R}$.

¹³See <http://nces.ed.gov/fastfacts/display.asp?id=569>.

To capture current tax policies, we use an approximation of Guner, Kaygusuz, and Ventura (2013) for effective marginal tax rates in the year 2000.¹⁴ We use the year 2000 because individuals in the NLSY97 are 18 in the year 2000 on average. Marginal tax rates concavely increase and converge to roughly 32%. We set the lump sum element of the tax code $\mathcal{T}(0)$ to minus \$1,800 a year. For average incomes this fits the deduction in the US-tax code quite well.¹⁵ For low incomes this reflects that individuals might receive transfers such as food stamps.¹⁶ We set the value of exogenous government spending to 11.2% of the GDP, which is the value that leads to a balanced government budget. This value is a bit low, but this should not be too surprising as we do not take into account corporate taxes or capital income taxes.

For tuition costs, we take average values for the year 2000 from Snyder and Hoffman (2001) for the regions Northeast, North Central, South and West, as they are defined in the NLSY. For all these regions we also take into account the amount of money coming from the taxpayer that is spent per student, which has to be taken into account for the fiscal externality. Both procedures are described in detail in Appendix A.1.1. The average values are \$7,434 for annual tuition and \$4,157 for the annual subsidy (public appropriations) per student. Besides these implicit subsidies, student receive explicit subsidies in the form of grants and tuition waivers. We estimate how this grant receipt varies with parental income and ability in Section A.1.4 using information provided in the NLSY. Finally, we make the assumption that individuals can only borrow through the public loan system. In the year 2000, the maximum amount for Stafford loans per student was \$23,000. The latter assumption does not seem innocuous. For our results about the desirability of increasing college subsidies, it is rather harmless because we show how our results can be understood in terms of sufficient-statistics and our quantified model predicts values for these sufficient statistics that are in line with empirical evidence.

4.3 Estimation of Wage Functions

As a first step, we estimate $G_{\theta}^{co}(w)$ and $G_{\theta}^{hs}(w)$. Our approach to estimate the relationship between innate ability, education and labor market outcomes relates to Abbott, Gallipoli, Meghir, and Violante (2013) and Johnson (2013).

In our model, y refers to an average income over the lifetime as we only have one labor market period. Assume that for a high school graduate i , one observes y_t^i for $t = 1, \dots, 48$ – i.e. from 18 to 65. The discounted present value of earnings (at age 18) is then given by $\sum_{t=0}^{48} \frac{y_t^i}{(1+r)^{t-1}}$. Simply taking the average over y_t to obtain the relevant income for our model

¹⁴We use the “Gouveia-Strauss”-specification including local sales taxes and take the average over all individuals. The parameters can be found in Table 12 of Guner, Kaygusuz, and Ventura (2013).

¹⁵Guner, Kaygusuz, and Ventura (2013) report a standard deduction of \$7,350 for couples that file jointly. For an average tax rate of 25% this deduction could be interpreted as a lump sum transfer of slightly more than \$1,800.

¹⁶The average amount of food stamps per eligible person was \$72 per month in the year 2000. Assuming a two person household gives roughly \$1,800 per year. Source: <http://www.fns.usda.gov/sites/default/files/pd/SNAPsummary.pdf>

would be misleading since discounting is not taken into account. Thus, we use annuitized income $\tilde{y}_i$ which is given by:

$$\tilde{y}_i = \frac{\sum_{t=1}^{48} \frac{y_t^i}{(1+r)^{t-1}}}{\sum_{t=1}^{48} \frac{1}{(1+r)^{t-1}}}.$$

For some more details on calculating $\tilde{y}_i$, see Appendix A.1.2. Everyone with less than 16 years of schooling is defined as a high school graduate.¹⁷ Everyone with 16 or more years of schooling is defined as a college graduate.

We run separate regressions, one for high school graduates and one for college graduates, of the form:

$$\ln \tilde{y}_i = \alpha_{ce} + \beta_e^{IN} \ln(AFQT_i) + \varepsilon_{ei}^{inc}, \quad (17)$$

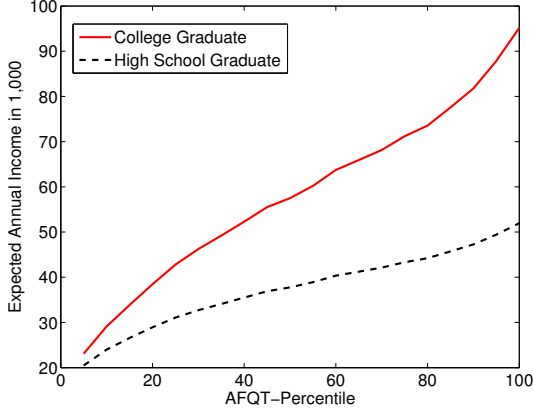
for $e = hs, co$. α_{ce} is a cohort-education fixed effect. We find that a one percent increase in AFQT-test scores leads to a 1.88% increase in income for college graduates and 1.28% increase in income for high school graduates, which reflects a complementarity between skills and education. This procedure gives us the mean of log incomes as a function of an individual's AFQT-score and education level. Based on that, we then calculate the respective average annual income over the life cycle for each AFQT-score and education level. We assume errors are normally distributed, so income is distributed log-normally. To determine the second moment of this log-normal distribution across education and innate ability levels, we use the sample variances of the error terms from (17) for each education level.

Top incomes are underrepresented in the NLSY as in most survey data sets. Following common practice in the optimal tax literature (Piketty and Saez 2013), we therefore append Pareto tails to each income distribution, starting at incomes of \$350,000. We set the shape parameter a of the Pareto distribution to 2 for all income distributions.¹⁸ Figure 1(a) shows the expected annual before tax income as a function of the AFQT (in percentiles) for both education levels and clearly demonstrates the complementarity between innate ability and education, which has also been highlighted in previous papers (Carneiro and Heckman 2005). The red bold line in Figure 1(b) shows how this translates into an expected NPV difference in lifetime earnings. As was argued in the theoretical section, the returns to education play an important role for the fiscal effects of an increase in college enrollment. The additional tax payment (again in NPV) is clearly increasing in AFQT (black dotted line). To get the overall impact on the

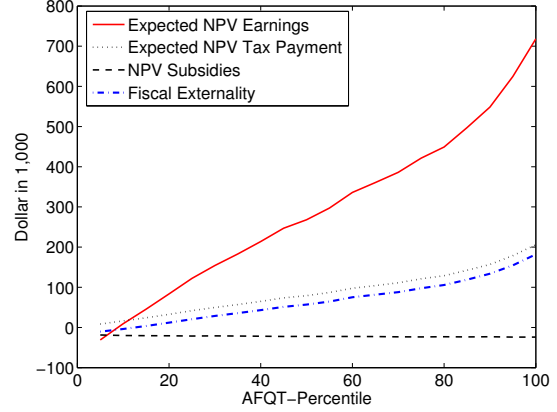
¹⁷Note that this definition also includes high school dropouts and individuals with community college degrees. We also worked with different specifications but our main results were not significantly affected.

¹⁸Diamond and Saez (2011) find that starting from $\approx \$350,000$ the Pareto parameter is constant and 1.5. Since their data are for 2005 and our data are also for earlier periods, we choose a Pareto parameter of 2 because top incomes were less concentrated earlier. The rationale for having the Pareto parameter independent of education and innate ability is that we estimated this parameter separately for each education group, obtaining estimates very close to 3 in both cases. Additionally, we tested for any systematic patterns for a direct relationship between this parameter and θ without finding any robust or significant patterns. The next consideration is the well-known problem that high incomes are severely under-sampled in survey data as the NLSY or the PSID. We choose 2 as explained in the beginning of this footnote.

government budget, subsidies have to be subtracted, which are given by the black dashed-line. Subsidies are increasing in ability which reflects the fact individuals with higher ability currently obtain higher scholarships (merit-based financial aid), which we elaborate in Section A.1.4. The net impact on public funds is given by the blue dashed-dotted line.



(a) Expected Annual Income



(b) NPV Income and Fiscal Externality

The last step consists of calibrating the respective skill/wage distribution from the income distributions by using the approximation of the US tax system from Guner, Kaygusuz, and Ventura (2013) and exploiting the first-order condition of individuals as pioneered by Saez (2001) and done in many papers after that.

4.4 Estimation of Psychic Costs

Based on the estimated reduced form relationships, we can calculate two value functions for each individual in the data: V_{co}^i and V_{hs}^i . These value functions are determined by innate ability and parental income.¹⁹ In line with the empirical literature, we assume that the decision to go to college is also influenced by heterogeneity in preferences for college. We assume that these psychic costs are determined by parental education and by innate ability – see Cunha, Heckman, and Navarro (2005), among others.²⁰ These assumptions give us a binary choice model

$$P(co_i = 1) = Prob(Y_i^* > 0)$$

¹⁹As parental transfers also depend on the variable *depkids*, we set the variable *depkids* to its sample mean when calculating the implied transfers for an individual as a function of parental income. The same is true for the variable *black* because it determines grant receipt. Given the correlation of *black* with parental income and AFQT, we do not take the overall sample mean but rather the respective share of blacks for each combination of parental income and AFQT.

²⁰The literature also suggests that individuals that grew up in urban areas are more likely to go college. The coefficient did not turn out as significant in our estimation and we therefore do not include it in our analysis. The inclusion of the variable does not affect any of our results.

where

$$Y_i^* = V_{co}^i - V_{hs}^i + \beta_1^{pc} + \beta_2^{pc} AFQT_i + \beta_3^{pc} S_i^{father} + \beta_4^{pc} S_i^{mother} + \varepsilon_i^{pc}$$

and where $\varepsilon_i^{pc} \sim N(0, \sigma)$. We restrict the coefficient on the difference in the value function to be one, as utility is our unit of measurement. For the power of the estimation, however, this is no restriction as we have one degree of freedom in parameter choice. As expected, all the variables have a positive and significant impact on the college choice, see Table 5 in the appendix.

Based on these estimations, we calculate the estimated psychic cost for each individual:

$$\hat{\kappa}_i = -\hat{\beta}_1^{pc} - \hat{\beta}_2^{pc} AFQT_i - \hat{\beta}_3^{pc} S_i^{father} - \hat{\beta}_4^{pc} S_i^{mother} - \hat{\varepsilon}_i^{pc}.$$

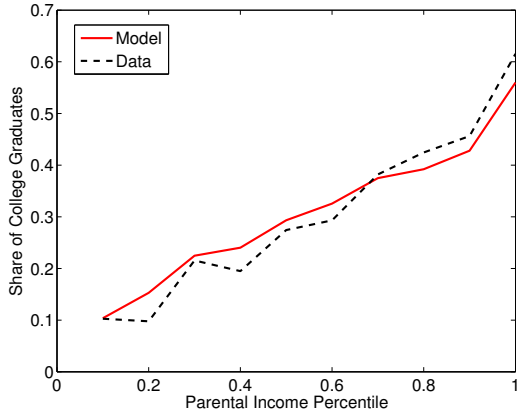
where $\hat{\varepsilon}_i^{pc} \sim N(0, \hat{\sigma})$. We draw 1,000 values for each ε_i and then fit a normal distribution of κ conditional on innate ability and parental income. Finally, we are then equipped with the joint distribution of parental income, innate ability and psychic costs.

4.5 Model Performance

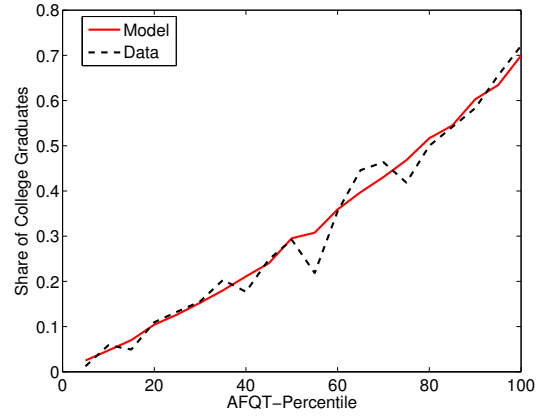
In order to assess the suitability of the model for policy analysis, we look at how well it replicates well known findings from the empirical literature and especially quasi-experimental studies. Note that the performance is also evaluated on moments which were *not* targeted in the estimation.

Graduation Shares. Figure 1 illustrates graduation rates as a function of parental income and AFQT in percentiles respectively. The bold lines indicate results from the model and the dashed lines are from the data. We slightly underestimate the parental income gradient. The correlation between AFQT and college graduation, however, is fitted well. The overall number of individuals with a bachelor degree is 30.56% in our sample and 30.95% in our model. Data from the United States Census Bureau are very similar: the share of individuals aged 25-29 in the year 2009 holding a bachelor degree is 30.6% – this comes very close to our data, where we look at cohorts born between 1980 and 1984.

Responsiveness of Graduation to Grant Increases. Many paper have analyzed the impact of increases in grants or decreases in tuition on college enrollment. Kane (2006) and Deming and Dynarski (2009) survey the literature. The estimated impacts of a \$1,000 increase in yearly grants (or a respective reduction in tuition) on enrollment ranges from 1-6 percentage points, depending on the policy reform and research design. Numbers differ since some of the evaluated programs were targeted towards low income groups and others were not, and sometimes the higher amount of grants was associated with a lot of paperwork, which might create selection. The majority of studies arrive at numbers between 3 and 5 percentage points,



(c) Graduation Rates and Parental Income



(d) Graduation Rates and AFQT

Figure 1: Removing Borrowing Constraints

however. As our model is a model of college graduation instead of college enrollment, the numbers are not directly comparable for two reasons: (i) not all of the newly enrolled students will indeed graduate with a bachelor's degree, (ii) some of the newly enrolled students enroll in community colleges and (iii) students that have enrolled also for lower grants are less likely to drop out of college. Relatively little is known about (iii). Concerning (i), we know that in the year 2000 roughly 66% of newly enrolled students enroll in 4-year institutions (Table 234 of Snyder and Dillow (2013)). Of those 66%, only slightly more than half should be expected to graduate with a bachelors degree. In Section A.2.2, we estimate that the dropout probability of the marginal students in our model is 45%. However, of those initially enrolled at two year colleges, also 10% graduate with a bachelors degree (Shapiro et al. (2012), Figure 6). Thus, translating the 3-5 percentage points increase in enrollment into numbers for graduation rates, we get 1.2-2 percentage points when taking into account (i) and (ii). Taking into account (iii) would yield slightly higher numbers, however, there is no strong empirical evidence about this effect that would guide us about the quantitative importance. In order to test our model, we increase the grant for each individual by \$1,000: bachelor degree completion rises by 1.48 percentage points. This number clearly lies within the reasonable range of estimates of 1.2-2.

A more recent study by Castleman and Long (2013) looks at the impact of grants targeted to low income children. Applying a regression-discontinuity design for need-based financial aid in Florida (Florida Student Access Grant), they find that a \$1,000 increase in yearly grants for children with parental income around \$30,000 increases enrollment by 2.5 percentage points. Interestingly, they find an even larger increase in the share of individuals that obtain a bachelor degree after 6 years by 3.5 percentage points. After 5 years the number is also quite high at 2.5 percentage points. These results show that grants can have substantial effects on student achievement conditional on enrollment.

Importance of Parental Income. It is a well known empirical fact that individuals with higher parental income are more likely to receive a college degree, see also Figure 1(c). However, it is not obvious whether this is primarily driven by parental income itself or variables correlated with parental income and college graduation. Using income tax data and a research design exploiting parental layoffs, Hilger (2014) finds that a \$1,000 increase in parental income leads to an increase in college enrollment of 0.43 percentage points. Using a similar back of the envelope calculation as in the previous paragraph – i.e. that a 1 percentage point enrollment increase leads to a 0.40 percentage points increase in graduation rates – this implies an increase in graduation rates of .17 percentage points. To test our model, we increased parental income for each individual by \$1,000 and obtained increases in bachelors completion by 0.10 percentage points. In line with Hilger (2014), our model predicts a small effect of parental income, even a bit smaller but in line with Hilger (2014).

The College Wage Premium and Marginal Returns. The college-earnings premium in our model is 100%, i.e. the average income of a college graduate is twice as high as the average income of a high-school graduate. As our earnings data are for the 1990s and the 2000s, this is well in line with empirical evidence in Oreopoulos and Petronijevic (2013); see also Lee, Lee, and Shin (2014). Doing the counterfactual experiment and asking how much the college graduates would earn if they had gone to college, we find that the returns to college are 64.40%. This implies a return of 12.43% for one year of schooling, which is in the upper half of the range of values found in Mincer equations (Card 1999, Oreopoulos and Petronijevic 2013).²¹

The more important number for our analysis is the return to college for *marginal students*. We find it to be slightly lower at 59.71%, which implies a return to one year of schooling of 11.71%. This reflects that marginal students are of lower ability on average than inframarginal students and also is in line with Oreopoulos and Petronijevic (2013). A clean way to infer returns for marginal students is found in Zimmerman (2014). In his study marginal refers to the *academically marginal* around a GPA admission cutoff. He finds returns of about 9.9% per year.²² However, his number refers to the *academically marginal* students (implying a GPA of 3), whereas in our thought experiment we refer to those students who are marginal w.r.t. to a small change in financial aid – these students are likely to be of higher ability than the academically marginal students. We explore this issue and make use of the fact that the NLSY also provides GPA data. In fact our model gives a return to college of 52.29% for students

²¹The calculation is as follows. In a Mincer regression, the log of earnings is regressed on years of schooling. The difference in $\log(1.64y)$ and $\log(y)$ is equal to $\log(1.64)$. Dividing by four years of schooling (for a bachelor degree) yields 12.43% per year of schooling.

²²He finds gains of 22% to obtain four-year college admission, which should be compared to the return of community colleges, which are the most frequent outside options for those students and take on average about 2 year less to complete. In addition, his findings are for earnings around 8 and 14 years after high school completion. Given that college students have a steeper earnings profile (see, e.g., Lee et al. 2014), these numbers are likely to underestimate the return to lifetime earnings.

with a GPA in the neighborhood of 3, which implies a Mincer return of 10.51% for one year of schooling – which comes very close to the 9.9% from Zimmerman (2014).

Working with a marginal treatment effect framework, Carneiro, Heckman, and Vytlačil (2011) estimate returns for marginal students with the NLSY79. They find smaller returns. Because both method and sample are different, it is not clear where precisely the differences are coming from. Motivated by the smaller returns in Carneiro, Heckman, and Vytlačil (2011), we provide robustness checks with calibrations matching their numbers.

Finally, we do not account for differing rates of unemployment and disability insurance rates. Both numbers are typically found to be only half as large for college graduates (See Oreopoulos and Petronijevic (2013) for unemployment and Laun and Wallenius (2013) for disability insurance). Further, the fiscal costs of Medicare are likely to be much lower for individuals with college degree. Lastly, we assume that all individuals work until 65 not taking into account that college graduates on average work longer (Laun and Wallenius 2013). These facts would strengthen the case for an increase in college subsidies.

The Role of Borrowing Constraints. To assess the importance of borrowing constraints, we completely remove them to ask by how much graduation increases. In this experiment enrollment increases by 4.10 percentage points from 30.95% to 35.05%. This value is in the realm of values the literature has found, see, e.g., Johnson (2013) and Navarro (2011). As Figure 2(a) reveals, the removal of borrowing constraints has larger effects for low income children. Figure 2(b) illustrates the importance of borrowing constraints for individuals with different innate abilities. Naturally, individuals with high ability have the strongest need for more borrowing because of high expected future earnings.

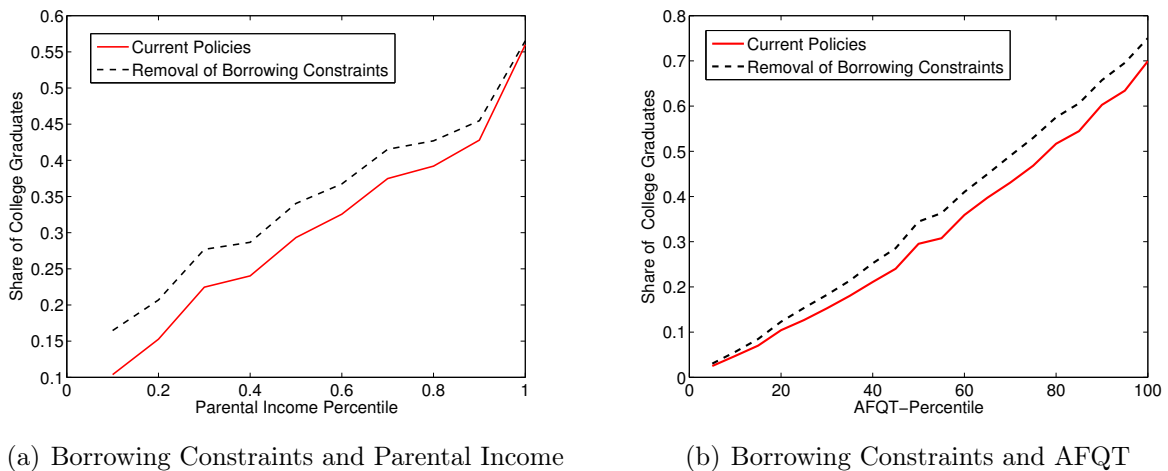


Figure 2: Removing Borrowing Constraints

5 Quantitative Results

5.1 Fiscal Effects of Reforms

5.1.1 Type Independent Policies

We first evaluate the fiscal effect of a marginal increase in education subsidies for every college student independent of characteristics. The three estimated parameters to calculate (10) are: the share of inframarginal students at 30.95%, the share of marginal students at .000381%, and the average fiscal externality per marginal student at \$70,940. This calculation reveals that $\gamma_R^G = -.13$. Thus, the return on one marginal dollar invested into education grants is -13%. In other words, an increase in universal college subsidies is close to being self-financing; for every marginal invested dollar, the taxpayer gets \$0.87 back. Thus, increasing subsidies is to a large degree self-financing.²³ It is straightforward to evaluate this formula for other values of these sufficient statistics. A lower responsiveness to a grant increase would make it smaller. It would increase for higher returns to education, a lower interest rate at which the government discounts and for a more progressive tax system. If one sets the real interest rate to 1% instead of 3%, for example, $\gamma^G = .46$ and increasing subsidies would be a free lunch. Finally, the lower the share of inframarginal students, the more profitable is a grant increase for the government.

The Role of Parental Transfers. Throughout our analysis we assume that parental transfers are policy invariant. In a large scale OLG model Abbott, Gallipoli, Meghir, and Violante (2013) argue that each dollar of government transfers to students crowds out 20-30 Cents of parental transfers on average. What would that imply for our results above? First of all, the fact that only 70-80 Cents would reach the inframarginal students would not affect our calculations at all because this would not affect their behavior. How about the marginal students? It is simple to show that the endogeneity of parental transfers with respect to policies would *not affect* the share of marginal students in our setting. The reason is that parental transfers are contingent on the education choice. Therefore, whenever there is no commitment problem of the parents in the sense that they provide a different amount of transfers than they announce before children make their education decision, the set of marginal students is exactly the same as in the case of policy invariant parental transfers.

Robustness. Our analysis abstracts from two important features: (i) college dropout as well as (ii) general equilibrium effects on the college wage premium. In Appendix A.2.2, we show how our results are altered by taking into account dropout – in that case the government does

²³Holter (2014) studies the impact of public educational expenditure and income taxes on intergenerational earnings persistence and as a side result finds that implementing the Danish education subsidies in the US would be self-financing.

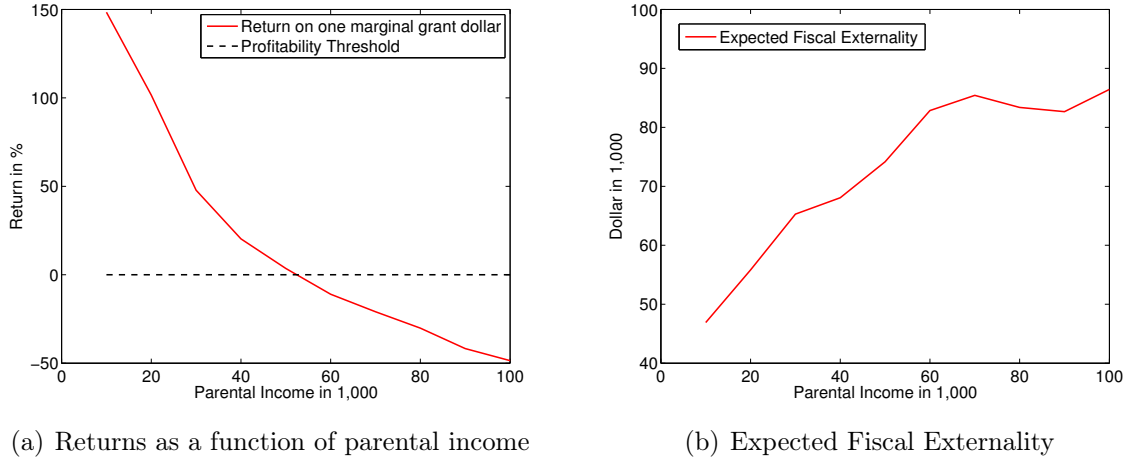


Figure 3: Grant Increase for Different Parental Income Groups

not only have to pay subsidies to those students who graduate but also to those that do not graduate. In that case an increase in $\mathcal{G}$ is less profitable than in the case without dropout. However, the effect is not huge: $\gamma^{\mathcal{G}}$ decreases from -.13 to -.30. Further, we are not taking into account that higher subsidies should also decrease the dropout rate as recent empirical evidence has shown (Angrist, Autor, Hudson, and Pallais 2015) and which would make the results more positive again. In Appendix A.2.3, we show how general equilibrium effects influence our results. In that case, a higher share of college graduates decreases wages for those with a college degree and increases wages for those without a college degree. As well explore in detail in the appendix, the second effect even seems to slightly dominate, so general equilibrium effects make the reform more likely to be self-financing.²⁴

5.1.2 Need-Based Grants

Our next experiment calculates this fiscal effects of increasing need-based aid for incomes up to \$50,000. We find $\gamma^{\mathcal{G}} = .37$ in this case. This striking result suggests very high returns to an expansion in targeted subsidies. To better understand what the mechanisms are, we calculate γ for many points of the parental income distribution. The reforms here are "pointwise" at some parental income $\mathcal{I}$, meaning that we only increase grants for those students who have parental income $\mathcal{I}$. This transparently illustrates the mechanisms at work. Figure 3(a) illustrates these experiments. The profitability is very high for low incomes and strictly decreasing in parental income. Increasing grants for individuals with parental income around \$15,000 has a very high return of 150%, so that a \$1 increase in subsidies is estimated to increase tax revenue by \$2.50.

²⁴These considerations imply that we are not looking at how individual college decisions might change in the presence of general-equilibrium effects. This is an important long-run questions that we leave for future research. See Abbott, Gallipoli, Meghir, and Violante (2013) for a discussion on the long-term effects.

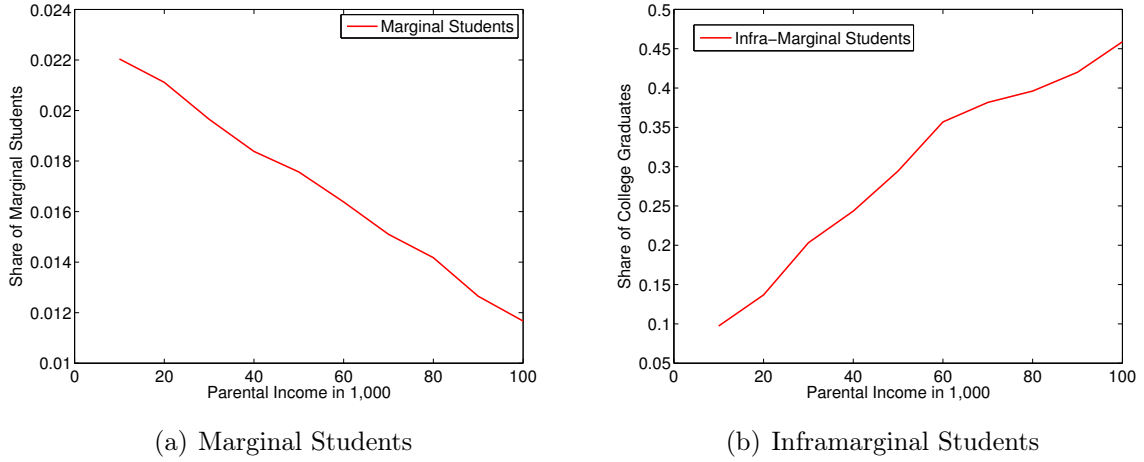


Figure 4: Fiscal trade-off of increase in subsidy

Returns are positive up to an income of slightly above \$50,000 and become negative afterwards. This is a result which might go against a first intuition. One could have expected that the returns to college are larger for children with a better parental background, as those kids are, arguably, more apt to take advantage of the opportunities offered by a college education. Figure 3(b) shows that this intuition is indeed correct in our quantified model because the expected fiscal externality is clearly increasing in parental income stemming from the positive correlation between θ and $\mathcal{I}$. However, this effect is counteracted and dominated by the fact that the share of marginal over inframarginal students (i.e. the semi-elasticity of enrollment) is heavily decreasing in parental income. As can be seen in Figures 4(a) and 4(b), the share of marginal students (normalized to a “marginal” annual grant increase of \$1,000) mildly decreases by a factor of roughly 2. Note that this is in line with Castleman and Long (2013) who find a strong responsiveness for children with low parental income. The share of inframarginal students varies more strongly and increases by a factor of roughly 4.5. Given that the fiscal costs of the reform are proportional to the share of inframarginal students, this implies that the marginal costs of subsidizing low income kids are 4.5 times lower than the costs of subsidizing high income children.

Our analysis shows that need-based education grants do not only favor social mobility and equity but are a good opportunity for the society to invest in future tax revenue. Remember that while the value of Δe for different income groups depends on our estimated model, the declining share of inframarginal students can be taken directly from the data. In fact, our results even *underestimate* the parental income gradient a bit, as discussed in Section 4.5. The fact that the result is to a large extent driven by those shares of inframarginal students across income groups should make the result particularly robust.

In the appendix, we extend the model to include parental labor supply incentives. The idea here is that need-based financial aid acts as an additional marginal tax rate on parents, which

will lead to a reduction of taxable income and therefore tax revenues. We find only very small effects of this extension on the desirability of need-based aid.

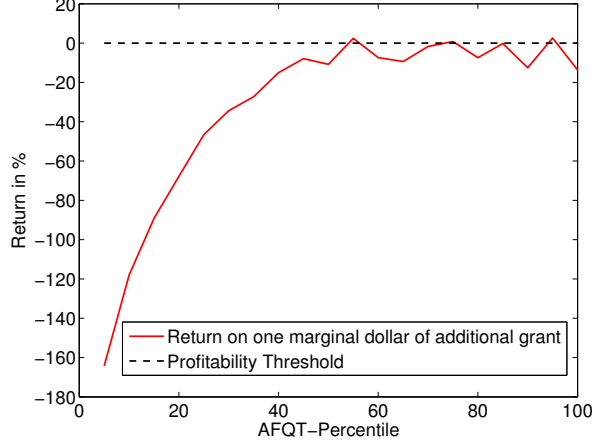


Figure 5: Returns as a function of AFQT

5.1.3 Merit-Based Grants

Another form of targeted student aid are merit-based grants. Figure 5 illustrates the return on one marginal dollar of aid as a function of the AFQT-score. As in Figure 3(a), we illustrate a hypothetical pointwise reform. Returns are positive if grant increases are targeted at individuals with AFQT-score above the 40th percentile. Afterwards it is almost flat. Two opposing effects cancel each other: whereas the fiscal externality is increasing in AFQT (see Figure 1(b)), the ratio of marginal over inframarginal students decreases.

5.2 Optimal Systems

We now quantify the optimal college subsidy 1, which does not condition on parental income or ability. Quantitatively, it will make a difference we include borrowing constraints or not. First, we assume that individuals can freely borrow. We allow the government to control two parameters: the lump sum element of the tax function and the college subsidy.

No Borrowing Constraints. In this case the social planner slightly increases the lump-sum element to \$2,005 from \$1,938, where the latter is the value of the transfer that the government can provide for current policies, but in the absence of borrowing constraints. Grants decrease from \$5,231 to \$3,905. The implied small decrease in enrollment is shown in Figure 6. The optimal Utilitarian graduation (holding marginal tax rates fixed) is 34.44%, and therefore slightly lower than the graduation rate given current policies absent borrowing constraints (35.05%). These results might seem at odds with our results in favor of higher subsidies

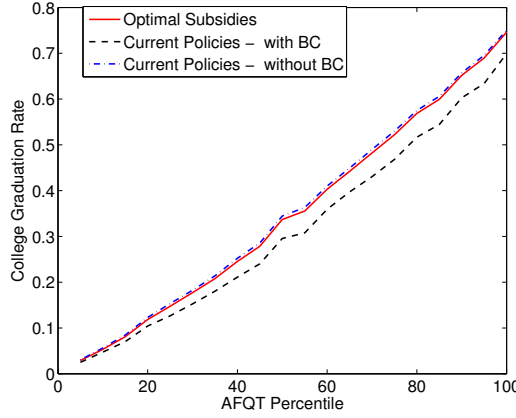


Figure 6: Optimal Education Subsidies for Given Income Taxes

in partial reform section. Why is the optimal subsidy here lower than the current level of subsidies? The reason is that we are considering a case where individuals can freely borrow. The additional effects of larger subsidies on enrollment are therefore significantly lower. Increasing subsidies would therefore require to lower the lump sum transfer that individuals receive (i.e. increase $T(0)$). A Utilitarian planner does the opposite at the margin since the social marginal utility of college students (given that they can freely borrow) is relatively high. To provide more consumption for individuals with low income realizations (hence, insurance), the planner slightly increases the lump-sum transfer that individuals receive at the cost of reducing the subsidy even though the reduction in subsidies comes with additional costs from the reduction in college graduation.

6 Optimal Taxes

6.1 Theory

We start by deriving a formula for the optimal marginal income tax rate in the spirit of Diamond (1998) and Saez (2001):²⁵

Proposition 2. *Optimal education independent marginal tax rates satisfy*

$$\frac{\mathcal{T}'(y(\omega^*))}{1 - \mathcal{T}'(y(\omega^*))} = \left(1 + \frac{1}{\varepsilon(\omega^*)}\right) \times \left(\frac{\int_{\omega^*}^{\bar{\omega}} (\beta^{co2} (1 - \mathcal{W}^{co}(\omega)) + \beta^{hs} (1 - \mathcal{W}^{hs}(\omega))) dG(\omega)}{(g^{co}(\omega^*)\beta^{co2} + g^{hs}(\omega^*)\beta^{hs}) \omega^*} \right. \quad (18)$$

$$\left. + \frac{\int_{\omega^*}^{\bar{\omega}} \int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \xi^{ind}(\theta, \mathcal{I}, \omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\mathcal{I}}(\theta) dF(\theta) d\omega}{(g^{co}(\omega^*)\beta^{co2} + g^{hs}(\omega^*)\beta^{hs}) \omega^*} \right)$$

²⁵In Appendix A.3.4, we also derive a formula for the optimal top tax rate.

where $\mathcal{W}^{co}(\omega)$ and $\mathcal{W}^{hs}(\omega)$ are as defined in (13) and (14). $\varepsilon(\omega)$ is the local labor supply elasticity and the college semi-elasticities $\xi^{ind}(\theta, \mathcal{I}, \omega)$ is as derived in Section 2.3.1.

Proof. See Appendix A.3.2 □

First, note that this formula holds for optimal as well as for suboptimal college subsidies. It differs from the seminal formula of Diamond (1998) in two respects. First of all, it is adjusted for period length and discounting. Second, the term

$$\int_{\omega^*}^{\bar{\omega}} \int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \xi^{ind}(\theta, \mathcal{I}, \omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\mathcal{I}}(\theta) dF(\theta) d\omega$$

shows up in the numerator. The formula is therefore related to the formulas of Saez (2002) and Jacquet, Lehmann, and Van der Linden (2013), where the extensive margin is due to labor market participation, or Lehmann, Simula, and Trannoy (2014) where the extensive margin captures migration. In these papers, the extensive margin is an unambiguous force towards lower marginal tax rates whenever workers pay more taxes than non-workers (or individuals that are on the margin of emigrating pay positive taxes). In contrast, the endogeneity of college enrollment does not necessarily lead to lower marginal tax rates as the additional term is ambiguous in its sign. First, we do not know the sign of $\Delta \mathcal{T}(\theta)$ in general. Second, we do not know whether higher taxes for individuals with $\omega > \omega^*$ indeed lead to lower college enrollment because of possibly counteracting income and substitution effects. Whether and to what extent the endogeneity of college enrollment leads to lower optimal marginal tax rates is, hence, a quantitative question.

6.2 Quantification

6.2.1 Optimal Taxes Given Education Subsidies

We consider the case where the government can set income taxes optimally given education subsidies as in the calibration. In Figure 7(a) optimal marginal tax rates are plotted as a function of yearly gross income. The shape of marginal tax rates is U-shaped, as is typically the case with exogenous education (Diamond 1998, Saez 2001). In order to illustrate the impact of the endogeneity of college enrollment, we also calculate a so called *self-confirming policy equilibrium* (SCPE) as defined in Rothschild and Scheuer (2013,14). Here a government is considered that naively sets policies under the assumption that college enrollment does not react to policies. A SCPE is defined as a situation where the college decisions implied by government's policies coincide with the expectation of the naive government. As can be seen in Figure 7(a), marginal tax rates are slightly higher in this case – the maximum difference is 1.2 percentage points. For top income tax rates, the endogeneity of enrollment plays no significant role.

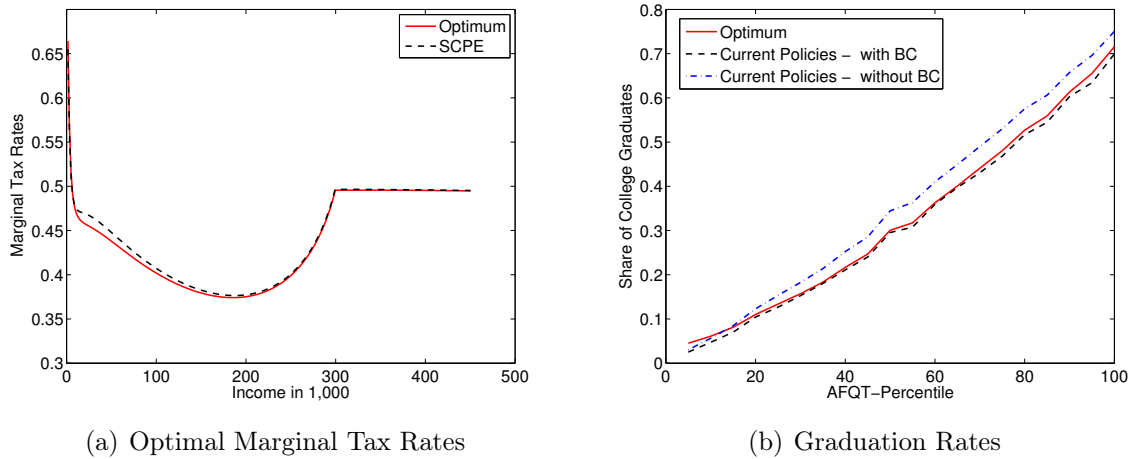


Figure 7: Optimal Income Taxes for Given Education Subsidies

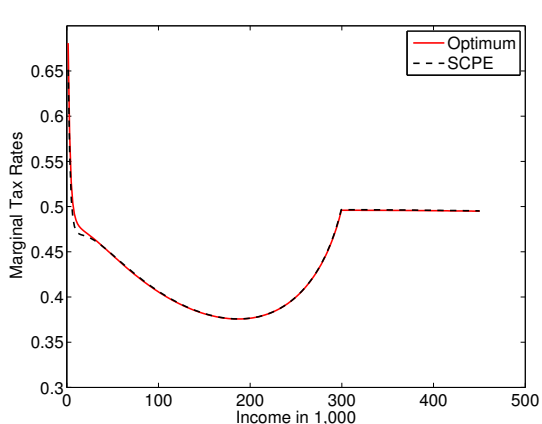
The share of college graduates in this case is 31.88% – this number should be compared to 35.05% enrollment in the absence of borrowing constraints. Intuitively, the Utilitarian planner uses the tax system to redistribute leading to higher taxes. As the planner cannot use subsidies to offset the disincentive effects from labor taxes in this scenario, the higher level of redistribution necessarily implies a decrease in enrollment. Interestingly, however, enrollment/graduation increases for low ability individuals. The much higher lump sum transfer, which is now at \$12,738, implies more insurance against low earnings. This is particularly relevant for low AFQT levels, as these individuals are particularly likely to earn low income even with a college degree. The additional amount of insurance allows individuals to fail and therefore encourages them to go to college. This effect is similar as in the theory of the welfare state by Sinn (1995), who argues that insurance achieved by redistribution can promote risky but socially desirable investment if individuals are risk averse.

6.2.2 Joint Optimal Income Taxes and Education Subsidies

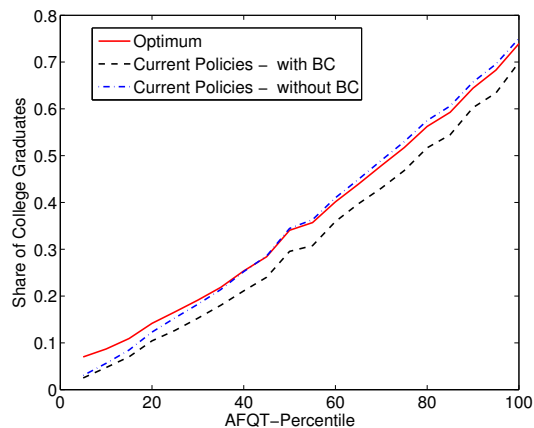
We now explore the case where the government can set income taxes and education subsidies jointly optimal. From a theoretical point of view, optimal subsidies are still characterized by the formula in Proposition 1 and optimal marginal tax rates by the formula in Proposition 3.

Optimal marginal tax rates are displayed in Figure 8(a). They are slightly higher as compared to the previous subsection. This result is intuitive as higher education subsidies can counteract tax distortions and therefore allow for more redistribution and social insurance.

In this scenario, we allow the government also to set college subsidies optimally. The government sets optimal education subsidies to \$16,957. This number is much higher than the number in Section 3.2.1, where we calculated the optimal college subsidy taking the current tax system as given. This is intuitive. Since optimal Utilitarian marginal tax rates are much



(a) Optimal Marginal Tax Rates



(b) Graduation Rates

Figure 8: Optimal Income Taxes and College Subsidies

higher than the ones in the current tax system, the fiscal externality imposed by the decision to go to college are higher which in turn increases the gains from subsidizing college. In addition, higher tax rates allow for more public spending on education.

This also explains why there is not a big difference between optimal marginal tax rates and those in an SCPE. The new term in Proposition 3 is now smaller. Since subsidies are at much higher levels (about 3-4 times the level), the fiscal externality is much smaller.

Lastly, as Figure 8(b) reveals, optimal education subsidies boost college enrollment in particular for lower AFQT-levels far above the numbers in the previous subsection, where subsidies were fixed. Also compared to current policies without borrowing constraints, graduation slightly increases from 35.05% to 35.25%.

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A Appendix

A.1 Appendix for Section 4

A.1.1 Tuition Fees and Public Costs of Colleges

First, we categorize the following 4 regions:

- Northeast: CT, ME, MA, NH, NJ, NY, PA, RI, VT
- North Central: IL, IN, IA, KS, MI, MN, MO, NE, OH, ND, SD, WI
- South: AL, AR, DE, DC, FL, GA, KY, LA, MD, MS, NC, OK, SC, TN , TX, VA, WV
- West: AK, AZ, CA, CO, HI, ID, MT, NV, NM, OR, UT, WA, WY

We base the following calculations on numbers presented by Snyder and Hoffman (2001). Table 313 of this report contains average tuition fees for four-year public and private universities. According to Table 173, 65% of all four-year college students went to public institutions, whereas 35% went to private institutions. For each state we can therefore calculate the average (weighted by the enrollment shares) tuition fee for a four-year college. We then use these numbers to calculate the average for each of the four regions, where we weigh the different states by their population size. We then arrive at numbers for yearly tuition & fees of \$9,435 (Northeast), \$7,646 (North Central), \$6,414 (South) and \$7,073 (West). For all individuals in the data with missing information about their state of residence, we chose a country wide population size weighted average of \$7,434.

Tuition revenue of colleges typically only covers a certain share of their expenditure. Figures 18 and 19 in Snyder and Hoffman (2001) illustrate by which sources public and private colleges finance cover their costs. Unfortunately no distinction between two and four-year colleges is available. From Figures 18 and 19 we then infer how many dollars of public appropriations are spent for each dollar of tuition. Many of these public appropriations are also used to finance graduate students. It is unlikely that the marginal public appropriation for a bachelor student therefore equals the average public appropriation at a college given that costs for graduate students are higher. To solve this issue, we focus on institutions “that primarily focus on undergraduate education” as defined in Table 345. Lastly, to avoid double counting of grants and fee waivers, we exclude them from the calculation as we directly use the detailed individual data about financial aid receipt from the NLSY (see Section A.1.4). Based on these calculations we arrive at marginal public appropriations of \$5,485 (Northeast), \$4,514 (North Central), \$3,558 (South), \$3,604 (West) and \$4,157 (No information about region).

A.1.2 Details on Income Regressions

For most individuals, we do not have information in every year. First of all, we never have information after age 53. Second, since 1994 the survey is conducted biannually. Third, we often have to deal with missing values. To resolve the first issue, we assume that incomes are flat afterwards, which is roughly what one finds in data sets with information on earnings over the whole life cycle. See, e.g., Figures 13 and 14 in Lee, Lee, and Shin (2014). Concerning the second issue, we take the average of the income in the year before and after. Concerning the third issue, we proceed similarly but also take values that are two and three years away if information for the year before and after is missing as well. All other years that are still missing are then just not taken into account for calculating $\tilde{y}_i$. Assume, e.g., that only income at age 19, 33 and 46 were observable. Then we would calculate

$$\tilde{y}_i = \frac{\frac{y_{i19}}{(1+r)} + \frac{y_{i33}}{(1+r)^{14}} + \frac{y_{i46}}{(1+r)^{27}}}{\frac{1}{(1+r)} + \frac{1}{(1+r)^{14}} + \frac{1}{(1+r)^{27}}}.$$

All incomes are measured in 2000 dollars.

Our estimates for the slopes are. $\hat{\beta}_{co}^{IN} = 1.88$ (0.186) and $\hat{\beta}_{hs}^{IN} = 1.28$ (0.074). As described in the main text, the second-moments of the log-normal parts are education dependent, so that until 350k, $\ln y$ is normal with standard deviation σ^e . We directly take the estimates for σ^e from the distribution of residuals from (17). The values are 0.6548 for college and 0.6631 for high-school.

A.1.3 Parental Transfers

In the NLSY97 we can observe the amount of transfers an individual obtained from its parents as well as family income. We take the constructed variable for parental transfers from Johnson (2013), who also takes into account the value of living at home as part of the parental transfer, for those individuals who cohabitate with their parents. We take yearly averages of those transfers for the ages 19-23. The sample average is \$6,703. We estimate the following equation:²⁶

$$\log(tr_i) = \alpha^{tr} + \beta_1^{tr} \log(\mathcal{I}_i) + \beta_2^{tr} co_i + \beta_3^{tr} depkids_i + \varepsilon_i^{tr}, \quad (19)$$

where *depkids* is the number of dependent kids living in the household of the parents. The coefficients are provided in Table 2. A 1% increase in parental income increases parental transfers by 0.31% and college graduates receive transfers that are 79% ($\exp(.5829) - 1$) higher than for high school graduates. Note that this implies that the absolute increase of parental transfers because of going to college is higher for high income kids. Johnson (2013) and Winter

²⁶We also estimated models with an interaction term between log parental income and college graduation. The coefficient on the interaction term is statistically insignificant.

Table 2: Transfer Equation

	Parental Income	College	Dependent Children
Coefficient	.3136***	.5829***	-.0667**
Standard Error	(.0449)	(.0563)	(.0329)

N=3,238. Robust standard errors. * $p \leq 0.10$, ** $p \leq 0.05$, *** $p \leq 0.01$.

(2014) have argued that it is crucial to take this effect into account to explain the large impact of parental income on college enrollment and completion.

Besides transfers that individuals receive during that time, they can also have some assets when they decide to study. In the NLSY97, information is provided on individual net worth at age 20. Certainly, this is not the best number for our purposes since it is highly influenced by choices at ages 18 and 19. We nevertheless take this noisy measure into account because it gives our quantitative model a better fit concerning the importance of parental income. To measure how net wealth varies with parental income, we estimate the following regression:

$$w_i = \alpha^w + \beta^w \mathcal{I}_i + \varepsilon_i^w. \quad (20)$$

We find a gradient for parental income of .127 (0.02) and an intercept of \$7,950 (1164). To obtain the parental transfer for the model, we take the implied parental transfer from equation (19) and adjust it by the implied level of wealth from equation (20) and thereby recalculate the wealth into an annual transfer.

A.1.4 Estimation of Grant Receipt

In practice, grants and tuition subsidies are provided by a variety of different institutions. Pell grants, for example, are provided by the federal government. In addition, there exist various state and university programs. To make progress, similar to Johnson (2013) and others, we go on to estimate the probability of grant receipt and the expected sum of the yearly grant directly from the data. Since many students do not receive any grants at all, we first estimate a binary model of whether an individual receives a grant.

$$P(\mathcal{G}_i > 0) = Prob(Y_i^* > 0)$$

where

$$Y_i^* = \alpha^{gr1} + \beta_1^{gr1} \mathcal{I}_i + \beta_2^{gr1} black_i + \varepsilon_i^{gr1}.$$

Interestingly, neither family size nor ability have a statistically significant effect at the extensive margin of grant receipt. As Table 3 reveals, blacks and low income children are more likely to obtain grants which captures affirmative action and need-based grants.

Next, we estimate the amount of grants conditional on receiving grants:

Table 3: Probit Estimation of Grants

	Parental Income	Black
Coefficient	-4.47e-06 ***	.2804***
Standard Error	(4.81e-07)	(.0623)

N=3,897. * p ≤ 0.10, ** p ≤ 0.05, *** p ≤ 0.01.

$$\log(gr_i) = \alpha^{gr2} + \beta_1^{gr2} \log(\mathcal{I}_i) + \beta_2^{gr2} black_i + \beta_3^{gr2} AFQT_i + \varepsilon_i^{gr2}. \quad (21)$$

Interestingly, family size is again not statistically significant. Besides grant generosity being need-based and in favor of blacks, generosity is also merit-based as $\hat{\beta}_3^{gr2} > 0$. The implied average grant is \$1,153 per year academic year. To calibrate the grant for an individual in our model we then multiply the implied grant from (21) with the probability of grant receipt implied by the probit equation of grant receipt. Finally we add the level of public appropriations – i.e. the level of subsidies that colleges receive from the government – that we obtain as described in Section 4.2.

Table 4: OLS for Grants

	Parental Income	Black	AFQT
Coefficient	-.2405***	.3709 ***	2.1203 ***
Standard Error	(0.0586)	(0.1143)	(0.3493)

N=968. * p ≤ 0.10, ** p ≤ 0.05, *** p ≤ 0.01.

A.1.5 Probit Estimates

Table 5: Probit Estimation of College Graduation

	AFQT	Father's Education	Mother's Education
Coefficient	.0140***	0.0832***	0.0504***
Standard Error	(.0011)	(.0097)	(.0105)

N=3,897. * p ≤ 0.10, ** p ≤ 0.05, *** p ≤ 0.01.

A.2 Appendix for Section 5.1

A.2.1 Need-Based Grant Reforms and Parental Incentives

Assume that parents have preferences of the form:

$$u(c, K, \mathcal{I}),$$

where K is total resources available for their kids. For the child's resources, we have $K = tr_{co} + \mathcal{G}(\theta, \mathcal{I})$. Parents choose their transfer tr_{hs} . The budget constraint of the parents is $c = \mathcal{I} - T(\mathcal{I}) - tr_{co}$. Their problem then reads as:

$$\max_{tr_{co}, \mathcal{I}} u(y - T(\mathcal{I}) - tr_{co}, tr_{co} + \mathcal{G}(\theta, \mathcal{I}), \mathcal{I}).$$

The first-order conditions are

$$u_K = u_c$$

and

$$u_c \cdot [1 - T'(\mathcal{I})] + u_K \cdot \left[\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}} \right] = -u_{\mathcal{I}}$$

Combining the first-order conditions yields:

$$u_c \cdot \left[1 - T'(\mathcal{I}) + \frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}} \right] = -u_{\mathcal{I}}$$

which implies that the parents will react to an decrease in $\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}$ in the same way as to an increase in $\mathcal{T}'(\mathcal{I})$. Denote by $\varepsilon_{\mathcal{I}^*, 1-T'+\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}}$ the elasticity of income with respect to an increase of the effective marginal tax rate.

The reform we are considering is to increase grants for children with parental income below a certain threshold $\mathcal{I}^*$ by $d\mathcal{G}$. This implies an increase of $\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}$ within a small interval $[\mathcal{I}^*, \mathcal{I}^* + \Delta \mathcal{I}]$ by

$$\Delta \left(\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}} \right) = \frac{d\mathcal{G}}{\Delta \mathcal{I}}$$

.

The impact of the change in parents behavior on public funds is then given by:

$$\left(\mathcal{T}' - \frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}} \right) \varepsilon_{\mathcal{I}^*, 1-T'+\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}} \frac{\mathcal{I}^*}{1 - \mathcal{T}' + \frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}} \frac{d\mathcal{G}}{\Delta \mathcal{I}}.$$

The mass of affected parents is $h_{\mathcal{I}^*}^{co} \Delta \mathcal{I}$, where $h_{\mathcal{I}^*}^{co}$ is the density of parents with income $\mathcal{I}^*$ and children that go to college. The overall effect is therefore

$$\left(\mathcal{T}' - \frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}} \right) \varepsilon_{\mathcal{I}^*, 1-T'+\frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}} \frac{\mathcal{I}^*}{1 - \mathcal{T}' + \frac{\partial \mathcal{G}(\theta, \mathcal{I})}{\partial \mathcal{I}}} d\mathcal{G}.$$

We set $\varepsilon_{\mathcal{I}^*, 1-T', \frac{\partial \mathcal{G}}{\partial \mathcal{I}}}^p = 0.33$. Since the relevant period where the parents could adjust their labor supply is only 4 periods, we choose a lower number than for the children, for whom we consider a lifetime labor supply decision.

Figure 9 illustrates the return on one dollar invested in grants for children with parental income below a certain threshold, taking into account parental incentives. The red bold line reflects the benchmark case (i.e. formula (11)) where parental incentives were not taken into

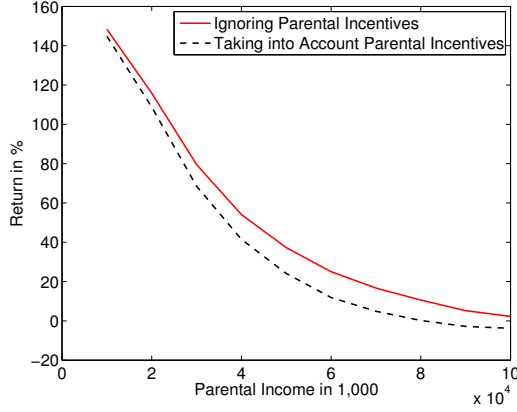


Figure 9: Need-based subsidy increase

account and the black dashed line reflects the case that they are taken into account. Returns are lowered, however, this effect is rather small. In particular, when the threshold is set at low parental income levels taking into account that parents might change their taxable income hardly affects the conclusions from the baseline case.

A.2.2 Robustness Check I: Dropout

In this subsection, we elaborate how our results change if college dropout is taken into account as well. We base our calculations on empirical evidence on dropout behavior as well as own estimates. Dropout changes the basic considerations from the previous subsections for two reasons: (i) subsidies are not only paid to those students who will graduate but also to those who drop out although for a shorter period of time. (ii) higher subsidies will not only induce individuals to study that will graduate but also some that will fail to graduate – whether this makes a subsidy increase fiscally more appealing depends on how large returns to ‘some college’ are.

We first look at effect (i) in isolation. Shapiro et al. (2012, Table 6) find that for the cohort which was first enrolled in a four year college in the fall of 2006, 62% graduated 6 years later. Thus, at most 38% never received a bachelor degree. Let’s first assume that all marginal students indeed graduate. How does a first back of the envelope calculation change our result? The adjusted formula looks as follows

$$\gamma_D^G = \frac{\Delta \bar{e} \times \Delta \bar{T}}{G^{co}(\bar{\omega}) \left(1 + \frac{1}{t_D} \frac{D_{infra}}{1 - D_{infra}}\right)} - 1 = -.38, \quad (22)$$

where $\frac{1}{t_d}$ adjust for the fact that dropouts do not receive subsidies over 4.5 years, but less. We assume that dropouts spend two years in college, which is the average number dropouts spent in college in the NLSY97. This analysis reveals that taking into account dropout is not unimportant. We did not take into account fact (ii) in this first calculation, i.e. the impact

Table 6: Dropout Probit

	$\mathcal{I}$	Father's Education	AFQT
Coefficient	-1.98e-06 ***	-.0622153 .***	-.0094686 ***
Standard Error	(6.47e-07)	(.0116132)	(.0013619)

N=1,921. * $p \leq 0.10$, ** $p \leq 0.05$, *** $p \leq 0.01$.

of marginal students that drop out on the government budget. It is not obvious what value to take for the dropout probability of marginal students. In addition, marginal students also partly benefit from college. To tackle the first issue, we estimated the dropout probability in the NLSY97 as function of parental income, father's education and AFQT. The results are displayed in Table 6.

As expected, all these variables have a negative impact on dropping out of college.²⁷ Predicting the dropout probability of marginal and inframarginal students for our model yields a dropout probability of 36% for individuals with characteristics of our infa-marginal students and 45% for individuals that have characteristics of our marginal students. Thus the formula should now look like

$$\gamma_D^G = \frac{\Delta \bar{e} \times \Delta \bar{\mathcal{T}} \left(\Delta \bar{\mathcal{T}} + \frac{D_{marg}}{1-D_{marg}} \Delta \bar{\mathcal{T}}^D \right)}{G^{co}(\bar{w}) \left(1 + \frac{1}{t_D} \frac{D_{infra}}{1-D_{infra}} \right)} - 1 = -.30 \quad (23)$$

A question that remains is how to quantify $\Delta \bar{\mathcal{T}}^{dropout}$. According to Lee, Lee, and Shin (2014) the earnings premium for 'some college' was between 25% and 40% between 1980 and 2005. Based on that, we assume that the earnings increase from some college (i.e. 2 years in college before dropout) is 30% of that from graduating.²⁸ For this case, we obtain $\gamma^G = -.30$ – thus even when taking dropout into account, the government gets back \$0.70 for each dollar invested.

Need-Based Grants We also investigate how our results for need-based grants change when taking into account college dropout. In particular, the results from Table 6 might raise doubt about the robustness of our result for the efficiency of need-based grants as low income children are more likely to drop out. We address this question and calculate the return on increasing subsidies for children with parental income below certain income levels again and take into account how the dropout probabilities of marginal and inframarginal students vary with income. Figure 10(a) shows how the dropout probability varies with parental income for marginal and

²⁷This is in line with reported statistics. Individuals with lower academic ability are more likely to dropout, see e.g. Chatterjee and Ionescu (2012). It has also been documented that parental characteristics play an important role, see e.g. Stinebrickner and Stinebrickner (2003).

²⁸The numbers cited from Lee, Lee, and Shin (2014) do not take into account different ability of those who graduate and those who dropout. The number that we take (i.e. 30%) is conditional on ability. Given that those who dropout are typically of lower ability, we consider our choice of 30% as a lower bound.

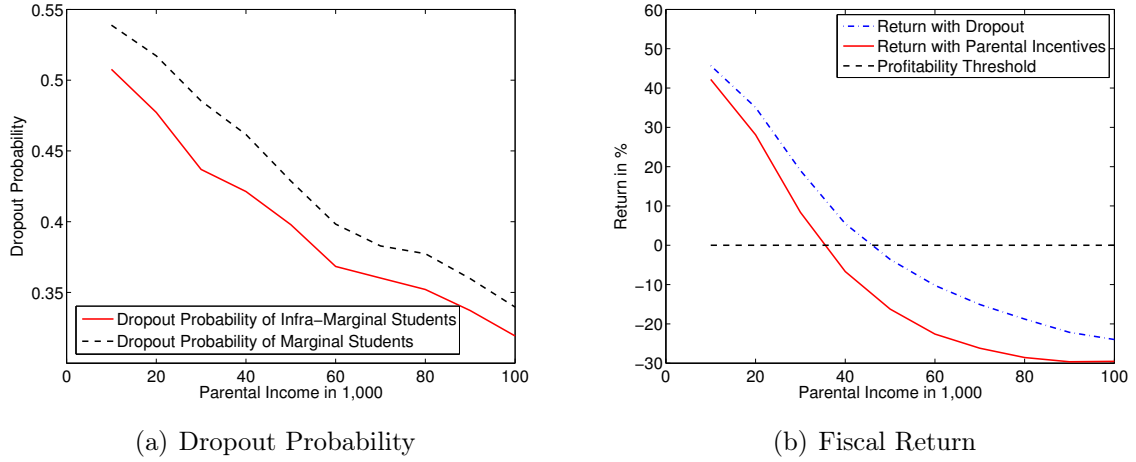


Figure 10: Revenue Gains as a Function of Annual Income

inframarginal students of our model. The dropout probability clearly decreases with parental income. As Figure 10(b) (which is the dropout-equivalent to Figure 9) shows, this has indeed an effect on how the profitability of grant increases varies with parental income (See blue dash-dotted line). However, the effect is not too large and does not change much with respect to the main conclusion. We even look at a ‘worst-case scenario’ and take parental incentives into account as well. As the red bold line reveals even in this scenario increasing subsidies for children with parental income below \$35,000 is a free lunch; this corresponds to those children in the lowest parental income tercile.

A.2.3 Robustness Check II: General Equilibrium

Our analysis so far abstracted from general equilibrium effects. A rising share of college graduates is likely to decrease the returns to college. What does this imply for our findings? A first educated guess might be that it weakens the case for an increase in subsidies. If returns to college decline, wages decline not only for the marginal but also for inframarginal students. It turns out, however, that for the reforms we consider, general equilibrium effects seem to strengthen the argument for increasing subsidies rather than weakening it. The reason is that the increase in college labor increases wages for high school graduates and therefore their contribution to public funds. There are approximately twice as many high-school than college graduates. As a consequence, the second effect (i.e. increases high-school wages) dominates and general equilibrium effects strengthen the case for an increase in subsidies.

Following common practice, we assume a CES production function of the following form:

$$F(H, L) = ((A_l L)^\rho + (A_h H)^\rho)^{\frac{1}{\rho}}$$

where $\rho \in (-\infty, 1]$ and $\sigma = \frac{1}{1-\rho}$ is the elasticity of substitution between college and non-college labor supply. For our model, we have

$$L = \int_{\Theta} \int_{\mathbb{R}_+} \int_{\Omega} y_{hs}(\omega) dG_{\theta}^{hs}(\omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta)$$

and

$$H = \int_{\Theta} \int_{\mathbb{R}_+} \int_{\Omega} y_{co}(\omega) dG_{\theta}^{co}(\omega) (1 - H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I}))) dK_{\theta}(\mathcal{I}) dF(\theta).$$

It is simple to show that wages per efficiency unit of high school and college labor are given by

$$w_{hs} = A_l^{\rho} \left(A_l^{\rho} + A_h^{\rho} \left(\frac{H}{L} \right)^{\rho} \right)^{\frac{1-\rho}{\rho}}$$

and

$$w_{co} = A_h^{\rho} \left(A_h^{\rho} + A_l^{\rho} \left(\frac{H}{L} \right)^{-\rho} \right)^{\frac{1-\rho}{\rho}}.$$

Without loss of generality, we set $w_{hs} = w_{co} = 1$ for our calibrated economy – it is w.l.o.g. since one cannot jointly identify the level of ω for workers and the wages for an efficiency unit of labor supply. We now repeat our analysis from above and ask what the revenue consequences are from marginally increasing grants for all (potential) college students. But now we assume that wages will change as a consequence of the increase in college labor supply. The consequence is that also tax payments of all individuals that do not change their college decision will change. The tax payments of college graduates will decrease as their wage decreases whereas the opposite is true for the high school graduates – for this simple exercise we assume that labor supply will be constant; as we argue below, endogenous labor supply would strengthen our results.

To compute how the number for $\gamma^{\mathcal{G}}$ changes, we therefore have to compute how the tax payments of those who do not change their behavior changes. Our adjusted formula thus reads as

$$\gamma_R^{\mathcal{G}} = \frac{\Delta \bar{e} \Delta \bar{\mathcal{T}} + G^{co}(\bar{\omega}) \Delta \bar{\mathcal{T}}_{GE}^{co} + G^{hs}(\bar{\omega}) \Delta \bar{\mathcal{T}}_{GE}^{hs}}{G^{co}(\bar{\omega})} - 1, \quad (24)$$

where $\Delta \bar{\mathcal{T}}_{GE}^{hs}$ and $\Delta \bar{\mathcal{T}}_{GE}^{co}$ reflect the average increase in the contribution to public funds due to general equilibrium effects on wages. We make the assumption that $\rho = .39$ which implies $\sigma = 1.64$ in accordance with Goldin and Katz (2009). We find that $\gamma^{\mathcal{G}} = -0.10$ for this case. Thus, general equilibrium effects increase this number by .03. Whereas tax payments from college graduates decrease, the opposite is true for high school graduates. As the share of high school graduates is larger than that of college graduates, it outweighs this effect even though high school wages increase less than college wages decrease and even though high school graduates face lower marginal tax rates given the progressivity of the tax code. Finally note that taking into account endogenous labor supply should strengthen our results: high school

graduates would work more and college graduates would work less. Given that the wage increase effect of high school graduates dominated the wage decrease effect, it is unlikely that the opposite is true for labor supply responses. To conclude, taking into account that relative wages may change because of an increase of college educated labor supply does not affect the fiscal externality argument strongly and is likely to strengthen it. Another important question that we leave for future research is how decisions of future generations to go to college will be affected by the implied decline in the college wage premium and what this implies for policies.

A.3 Theoretical Results: Appendix for Sections 3 and 6.1

A.3.1 Proof of Proposition 1

Formula (8) gives the impact on government revenue for a marginal increase in grants. Inserting (5) and (7) yields

$$\Gamma^G = d\mathcal{G} \times \int_{\Theta} \int_{\mathbb{R}_+} \Delta\mathcal{T}(\theta, \mathcal{I}) \xi^G(\theta, \mathcal{I}) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta) + d\mathcal{G} \times \beta^{co1} G^{co}(\bar{\omega}).$$

This expression does not take into account the impact on utility of students as the focus was solely on revenue. Taking into account utility as well and using the definition of the social welfare weights of students as defined in (15) changes the expression to

$$d\mathcal{G} \times \int_{\Theta} \int_{\mathbb{R}_+} \Delta\mathcal{T}(\theta, \mathcal{I}) \xi^G(\theta, \mathcal{I}) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\theta}(\mathcal{I}) dF(\theta) + d\mathcal{G} \times \beta^{co1} G^{co}(\bar{\omega}) (1 - \mathcal{W}^{st}).$$

Setting this expression to zero and using the standard definition of an elasticity gives the formula in Proposition 1.

A.3.2 Proof of Proposition 2

In an optimum:

$$\begin{aligned} d\mathcal{T}' dy \int_{\omega^*}^{\bar{\omega}} & \left(\beta^{co2} (G^{co}(\bar{\omega}) - G^{co}(\omega^*)) (1 - \mathcal{W}^{co}(\omega)) + \beta^{hs} (G^{hs}(\bar{\omega}) - G^{hs}(\omega^*)) (1 - \mathcal{W}^{hs}(\omega)) \right) dF(\omega) \\ & + CG(y(\omega^*)) + LS(y(\omega^*)). \end{aligned} \quad (25)$$

For a tax function to be optimal, this has to be zero. Setting the term to zero and solving for $\frac{\mathcal{T}'(y(\omega))}{1 - \mathcal{T}'(y(\omega))}$ yields Proposition 3.

A.3.3 Robustness Mirrleesian Income Taxation

Figure 11(a) shows that differences in marginal tax rates can get more substantial for lower values of the CRRA coefficient. The difference in optimal marginal tax rates and that of a

naive social planner differ up to 4.7 percentage points. By contrast, if the CRRA coefficient is higher than our benchmark, the difference gets negligible as Figure 11(b) reveals. These figures refer to the case where college subsidies are not set optimally but rather to their current level. If one looks at the case, where subsidies are optimal, the difference also becomes negligible for the case of a CRRA of 1, see Figure 12. Thus, in most cases the formula for optimal marginal tax rates of a planner that does not take into account the college margin are very similar to optimal marginal tax rates.

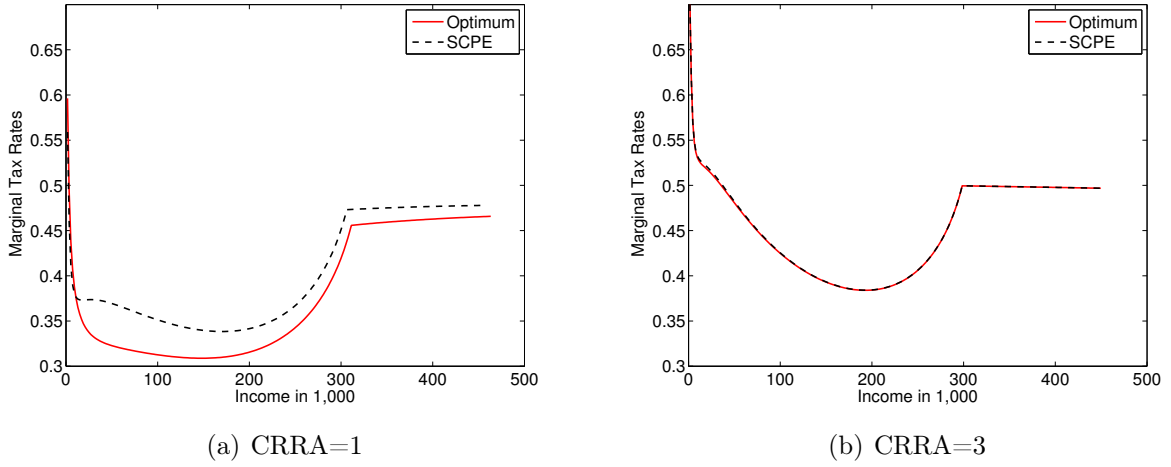


Figure 11: Robustness Optimal Income Taxes

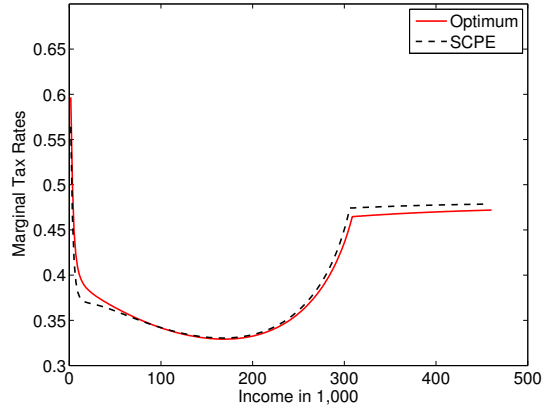


Figure 12: Robustness Optimal Income Taxes with Optimal Subsidies: CRRA=1

A.3.4 Optimal Top Tax Formula

If the upper tail of the earning distribution is Pareto distributed, one can derive a simple formula for the optimal top tax rate. Consistent with the empirical distributions we use, we

assume that for both education groups above a high-income threshold y^T , earnings are Pareto distributed with common parameter a .

Proposition 3. *Assume that welfare weights for top incomes converge to a constant $\overline{W}$. Then the optimal top marginal tax rate above income level y^T satisfies:*

$$\frac{\tau}{1-\tau} = \frac{1 - \overline{W} + \overline{CG} \cdot a}{1 + \epsilon \cdot a},$$

where $\overline{CG} = \frac{\int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \int_{\omega^T}^{\overline{\omega}} \xi^{ind}(\theta, \mathcal{I}, \omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\mathcal{I}}(\theta) dF(\theta)}{(\beta^{co2} + \beta^{hs}) y_m}$ and ω^T is the skill level associated with the threshold income level y^T and y_m is average income in the top bracket.

Proof. Mechanical effects by a small reform of τ are given by $\overline{M} = (\beta^{co2} + \beta^{hs})(y_m - y^t)(1 - \overline{W})$. Labor supply effects are $\overline{LS} = -\frac{\tau}{1-\tau} \epsilon y_m (\beta^{co2} + \beta^{hs})$. Using $\overline{LS} + \overline{M} + \overline{CG}(\beta^{co2} + \beta^{hs}) y_m = 0$ and $\frac{a}{a-1} = \frac{y_m}{y^t}$ gives the result. $\square$

$\overline{CG}$ captures the fiscal externality of the top tax rate through enrollment responses. When $\overline{CG}$ is zero, the formula collapse to the seminal Saez (2001) top tax formula. From the discussion in the theory section and the results from the quantified model, we expect $\overline{CG}$ to be negative as price (substitution) effects are likely to dominate. As a consequence, optimal top tax rates should be lower. This will confirmed next in the optimal policy simulations, although the effects of $\overline{CG}$ will be shown to be very small.

A.3.5 Optimal Education Dependent Taxes

Given that the government can observe the college choice, we now explore how optimal education dependent taxes should look like.

Proposition 4. *Optimal education dependent marginal tax rates satisfy*

$$\begin{aligned} \frac{\mathcal{T}^{co'}(y_{co}(\omega^*))}{1 - \mathcal{T}^{co'}(y_{co}(\omega^*))} &= \left(1 + \frac{1}{\epsilon^{co}(\omega^*)}\right) \times \left(\frac{\int_{\omega^*}^{\overline{\omega}} \beta^{co2} (1 - \mathcal{W}^{co}(\omega)) dG(\omega)}{g^{co}(\omega^*) \beta^{co2} \omega^*} \right. \\ &\quad \left. + \frac{\int_{\omega^*}^{\overline{\omega}} \int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \xi^{co}(\theta, \mathcal{I}, \omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\mathcal{I}}(\theta) dF(\theta) d\omega}{g^{co}(\omega^*) \beta^{co2} \omega^*} \right) \end{aligned} \quad (26)$$

and

$$\begin{aligned} \frac{\mathcal{T}^{hs'}(y_{hs}(\omega^*))}{1 - \mathcal{T}^{hs'}(y_{hs}(\omega^*))} &= \left(1 + \frac{1}{\epsilon^{hs}(\omega^*)}\right) \times \left(\frac{\int_{\omega^*}^{\overline{\omega}} \beta^{hs} (1 - \mathcal{W}^{hs}(\omega)) dG(\omega)}{g^{hs}(\omega^*) \beta^{hs} \omega^*} \right. \\ &\quad \left. + \frac{\int_{\omega^*}^{\overline{\omega}} \int_{\Theta} \int_{\mathbb{R}_+} \Delta \mathcal{T}(\theta, \mathcal{I}) \xi^{hs}(\theta, \mathcal{I}, \omega) H_{\theta, \mathcal{I}}(\tilde{\kappa}(\theta, \mathcal{I})) dK_{\mathcal{I}}(\theta) dF(\theta) d\omega}{g^{hs}(\omega^*) \beta^{hs} \omega^*} \right) \end{aligned} \quad (27)$$

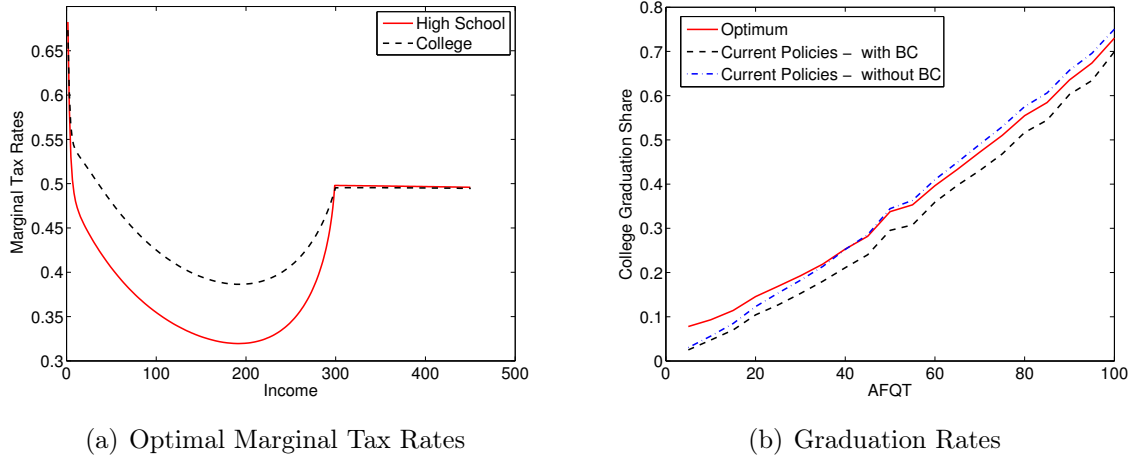


Figure 13: Optimal Education Dependent Taxes

These formulas are very similar to the education independent tax formula in Proposition 3. A difference is that all the terms now carry the superscript *co* or *hs*. It is well known from the tagging literature²⁹ that conditioning taxes on characteristics other than income is a powerful tool to tailor marginal tax rates to different subgroups. Importantly, education is not an exogenous tag here. Instead it is an endogenous choice. The planner therefore has to take into account the endogeneity of enrollment similarly as in Proposition 3. Whereas for education independent taxes, the effect of higher taxes on enrollment is theoretically ambiguous, this is not the case with education dependent taxes. Thus, for $\Delta\mathcal{T}(\theta) > 0$, endogeneity of enrollment is a clear force towards lower college taxes and higher high school taxes (as compared to the case with exogenous enrollment, i.e. classical tagging). Note that these formulas are related to Scheuer (2013), who considers the differential tax treatment of entrepreneurs and workers.

Quantitative Exercise. Figure 13(a) illustrates marginal tax rates as a function of gross income. Up to income levels of \$350,000 – where the Pareto tail starts for both distributions – marginal tax rates are higher for college graduates. For higher incomes, the result slightly reverses before marginal tax rates start to converge as the enrollment effect approaches zero. The optimal education subsidy is \$23,723. Overall graduation is now at 35.04%. Figure 13(b) illustrates graduation as a function of the AFQT-score.

²⁹This literature goes back to Akerlof (1978). For more recent contributions addressing this topic, see Mankiw and Weinzierl (2010), Cremer, Firouz, and Lozachmeur (2010) and Weinzierl (2014).