

THE EXPANSION OF COLLEGE EDUCATION IN THE UNITED STATES: IS THERE EVIDENCE OF DECLINING COHORT QUALITY?

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This article documents the expansion of college education in the United States and examines to what extent the increase in the number of college graduates may have led to a decline in the average quality of college graduates. Using the 1940–90 Census, we compare across birth year cohorts with varying levels of college completion. We find some weak evidence that college graduate men from highly educated cohorts earn a relatively smaller wage premium even controlling for the relative supply effect. However, these cohort quality effects account for only a small fraction of the recent fluctuation in the college wage premium. (JEL I20, J24, J31)

I. INTRODUCTION

In 1940, approximately 5% of men aged 20–64 were college graduates. By 1990, the fraction with a college degree had more than quadrupled to 22.6%. Conversely, more than 70% of men aged 20–64 had not finished high school in 1940. By 1990, the fraction of the male population who were high school dropouts had dipped below 20%. These changes imply that there has been an enormous increase in the supply of skilled workers in the U.S. economy over the past five decades. The magnitude of these changes also suggests that the composition of these groups, in terms of their underlying ability and talents, have also changed, making it difficult to compare these groups over time.

In this article we focus on the expansion of college education. We ask whether the increase in the number of college graduates over time lead to a decline in the average ability of college graduates. Following Becker (1967), we introduce a simple model of human capital accumulation with individuals who differ in

ability and borrowing costs. Under the assumption that the ability distribution is fixed and the more able attend college, we examine the conditions under which a rising share of college graduates leads to a decline in the average ability of the college-educated group.

The relationship between the expansion of higher education and the average ability of college attendees was explored by Taubman and Wales (1972) in their well-known study. Putting together evidence from a variety of test score studies, they reported that the average aptitude of those who attended college actually rose relative to those who did not attend college. We take an alternative, more aggregate approach in examining the expansion of college education and average ability in this article. Using the 1940–90 Census, we compare across cohorts with different levels of educational attainment. We ask whether college graduates from more educated cohorts receive a smaller college premium, even controlling for the relative supply effect. Thus we do not observe changes in ability directly but infer changes in ability through changes in relative wages.

There are several reasons why we view this alternative approach as useful. First, the test score studies are based on different tests administered in different states, which raises

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ABBREVIATIONS

PUMS: Public Use Micro Samples
CPS: Current Population Survey

questions about the comparability of the results over time. Second, the test scores are often available only for those who finished high school. Because the selection of high school graduates is also changing, it is difficult to infer changes in ability of college graduates. Third, test scores typically have little explanatory power in earnings regressions, suggesting that other dimensions of ability are important for earnings determination. Finally, many of these studies refer to the first half of the twentieth century, whereas we extend the analysis to the more recent period.

A number of recent papers (Murphy and Welch 1992; Katz and Murphy 1992; Topel 1997; Autor et al. 1998) have argued that the rapid increase in the share college graduates depressed college premiums in the 1970s. Welch (1979), Berger (1985) and Card and Lemieux (2000) show that these declines in college premiums were particularly associated with the Baby Boom cohorts who entered college in record numbers. In this article we examine to what extent these cohort effects may reflect a decline in quality associated with an accelerated pace of college attendance.

To preview our results, we find a small and marginally significant cohort quality effect that suggests that a 10% increase in the cohort-specific college share leads to a 0.6% decline in the college premium, holding everything else constant. These results suggest that changes in cohort quality played a minimal role in the recent period, with relative supply and demand shocks accounting for most (87%) of the variation in relative wages. These small relative wage effects related to changes in ability are in line with evidence from test scores reported in Murnane et al. (1995).¹

Section II begins by describing the aggregate changes in educational attainment over the period 1940–90. Although we concentrate mostly on prime-age men in this work, we also report changes in educational attainment of women. Section III presents our human capital framework. Section IV presents the results

from our cohort-level regressions. Section V presents a summary of our main findings.

II. THE EXPANSION OF COLLEGE EDUCATION IN THE UNITED STATES

One of the most notable changes in the U.S. labor market in the past several decades has been the increase in educational attainment of its labor force. In this section we document this change using the 1/100 sample of the 1940–90 Public Use Micro Samples (PUMS). To describe changes in the education distribution over time, we use an inclusive sample of men and women who were 20–64 years old. Our regressions results are based on men who are older, 25–64 years old. We restrict our analysis to the older group because they are more likely to have completed their education, and we examine men to ensure consistent labor market attachment. To calculate labor quantities, we include men 25–64 years old who were employed during the survey week but did not work in agriculture. To calculate wages, we chose a more select sample of these men who were wage and salary workers, worked full-time, worked at least 40 weeks, and earned at least one-half the legal federal minimum weekly wage. Our wage measure is the weekly wage calculated as annual earnings divided by weeks worked. Annual earnings were deflated using the personal consumption expenditure deflator from the national product and income accounts.

Table 1 reports shares of men and women of prime working age (20–64 years old) in four education categories: less than high school, high school graduate, some college, and college graduate. In 1940, as much as 74% of working-age men had less than a high school degree, and 5.4% had a college degree. By 1990, less than 20% were high school dropouts and 22.6% college graduates.² Table 1 also reports the educational attainment of women. The table shows that there has been an equally dramatic increase in the educational attainment for women, although women are still slightly less likely to be college graduates than

1. Our quality effects are smaller than those reported by Rosenbaum (1999), who examines the rise in the college premium with and without ability controls. Both studies use cohort-specific educational attainments to proxy for ability changes, but the difference in our results may be due to the fact that we focus exclusively on changes in ability *composition* while Rosenbaum also allows for changes in *returns* to ability over time.

2. The 1990 Census questionnaire changed emphasis from grades attended to degree obtained. To make the 1990 data comparable to the earlier years, we imputed the distribution of grades completed by reported education on the 1990 Census using the matched March samples for 1991–92. Details of this imputation procedure are available from the authors.

TABLE 1
Education Distribution 1940–90

Years of Schooling	1940	1950	1960	1970	1980	1990
Men						
< 12	73.6	63.3	54.5	39.7	25.4	18.6
=12	15.3	21.2	25.1	32.1	35.1	35.7
13–15	5.8	8.3	10.2	13.9	19.3	23.1
16+	5.4	7.3	10.2	14.3	20.2	22.6
Women						
< 12	70.0	59.0	51.1	38.3	25.4	17.7
=12	19.7	27.4	32.8	39.7	42.2	39.7
13–15	6.5	8.4	10.0	13.2	18.7	24.1
16+	3.8	5.2	6.1	8.8	13.6	18.6

Source: 1940–90 PUMS (1% samples). The samples include men and women aged 20–64.

the men. For example, even in 1990, the fraction of women who are college graduates is 18.6% compared to 22.6% for men. However, as will be illustrated later, the educational attainment of the most recent cohorts presents a different picture in that women are now more likely to be college graduates than their male counterparts.

Figure 1 and Figure 2 present educational attainment by birth-year cohort for men and women, respectively. The different panels present shares of the four education groups: less than high school, high school graduate, some college, and college graduate. The figures are based on data from the 1940–90 Census. As is apparent from the figures, men and women in any particular birth-year cohort acquire more schooling as they age. For example, a greater fraction of the cohort born in 1920 are college graduates in the 1950 Census (when the cohort is 30 years old) as compared to the 1940 Census (when the cohort was only 20 years old). To extrapolate the shares in each education category for the most recent cohorts, we run the following simple regression using cohort-level data:

$$(1) \quad e_{ct} = \alpha A_{ct} + \gamma_c + \mu_{at} + \varepsilon_{ct},$$

where e_{ct} is the share of the cohort in the education category, A_{ct} is a quartic in age, γ_c is the cohort specific effect measured by birth-year cohort dummies, μ_{at} are year-“young” interaction dummies where “young” is defined as 20–24 years old, and ε_{ct} is the error term.

We run these regressions for each of the four education groups and graph the shares predicted at age 35.

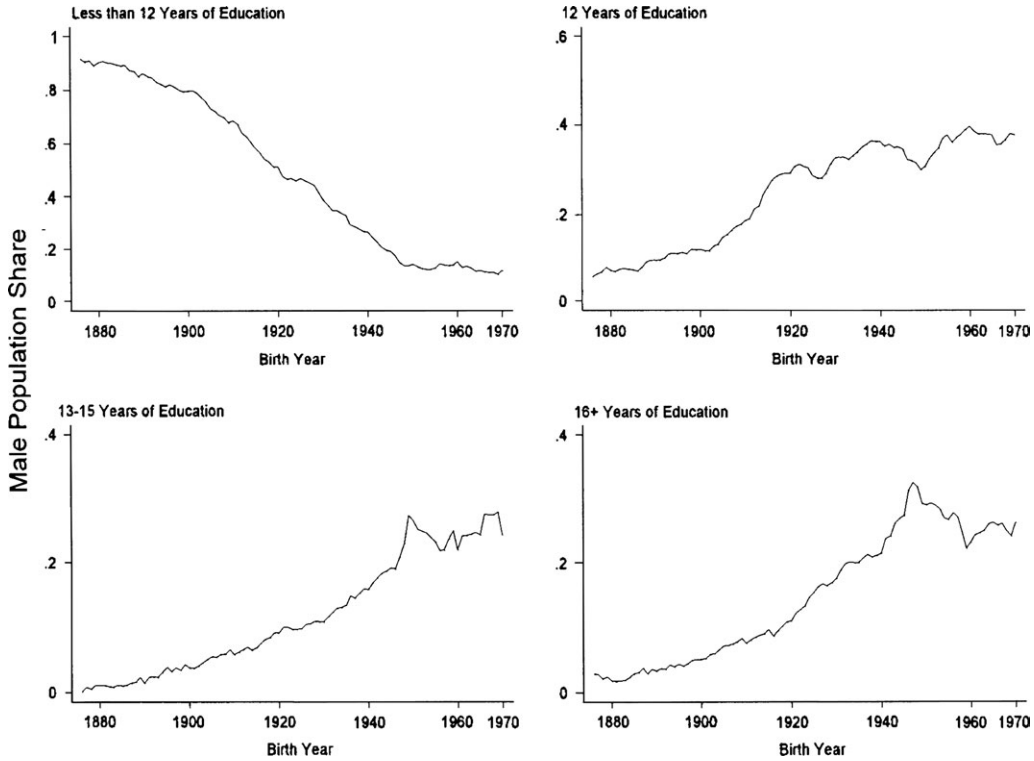
Figures 1 and 2 show that the fraction who had less than a high school degree fell dramatically from over 80% among the 1880 birth cohort to ~10% among the most recent cohorts. On the other hand, the fraction who are college graduates increased from less than 5% to over 30% among the most recent cohorts. For men, college graduation rates peaked with the 1948–50 birth cohorts (among whom approximately 31% were college graduates) and declined among the more recent cohorts. It is interesting that for women, college graduation rates began to accelerate starting with later cohorts, and there is little evidence of a decline among the more recent cohorts.

The trend in college enrollment in the most recent period is better illustrated in Figure 3 which is based on March Current Population Survey (CPS) data.³ The figure reports the fraction of 20–24-year-old men and women who are either college graduates or still enrolled in school during the survey years 1968–95. The left panel of Figure 3 shows that college enrollment peaked for men in the late 1960s with the cohorts born in the late 1940s, fell to ~24% in 1981, and has increased again in the most recent period, probably in response to the rise in the college premium. College enrollment among women (illustrated in the right panel of Figure 3) has more or less increased continuously throughout the 1970s and the 1980s and has surpassed the enrollment rates observed among men in the most recent period.

These figures illustrate the substantial variation in college completion rates across birth cohorts, particularly among men. The differences between men and women also suggest that factors such as the GI Bill during the 1940s and the deferment of the draft during the Vietnam War may have played important roles. It is these types of differences in college enrollments across birth cohorts that will be the basis of our empirical work in section IV.

3. Reporting of education on the March CPS switched between the 1991 and the 1992 surveys. We again used the March matched samples from 1991–92 to calculate the distribution of grades completed and school enrollment status by the new reported education variable.

FIGURE 1
Educational Attainment by Birth-Year
Cohort: Males



III. MODEL OF HUMAN CAPITAL ACCUMULATION AND COHORT QUALITY

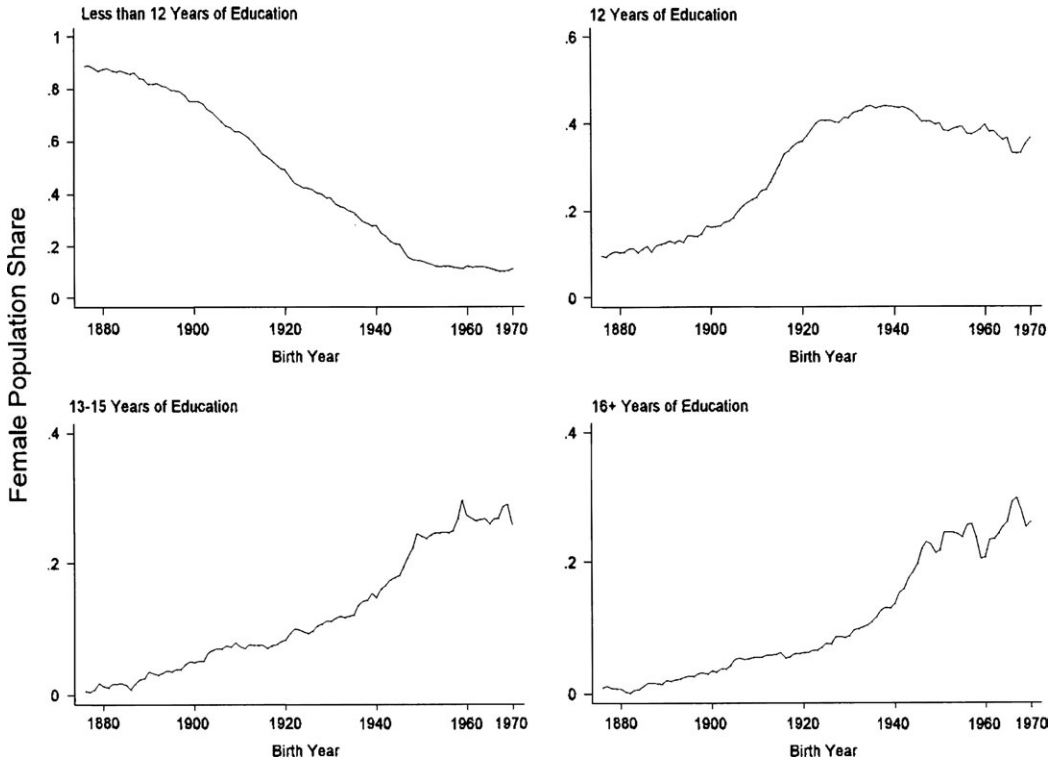
To provide a framework for discussion and for our empirical work, we examine a simplified version of the Becker (1967) model of human capital accumulation. This model has been reformulated by Rosen (1977) and generalized to incorporate heterogeneous human capital in Willis and Rosen (1979). Most recently it has been revived to discuss biases in estimating returns to schooling from cross-sectional data (Card 1994; Lang 1994). Our purpose here is to characterize, within the context of this well-known model, the conditions under which we can expect the size and average ability of the college educated group to negatively covary. The share of college educated can increase for a variety of reasons, such as an increase in price of skills acquired in college, decline in tuition costs, and decline in the cost of borrowing funds, which may be

affected by government loan or subsidy programs. In our discussion we consider these different cases.⁴

Individuals maximize the present value of lifetime earnings and compare the benefits and costs in deciding whether to obtain a college degree. The discounted benefit of college education net of forgone earnings, $R(\theta, r, \lambda)$, depends on the price of the skill acquired in college, θ , the interest (discount) rate, r , and the ability of the individual, λ . $R(\theta, r, \lambda)$ is increasing in θ , decreasing in r , and increasing in λ ($R_\theta > 0$, $R_r < 0$, and $R_\lambda > 0$). We denote the direct costs of college (such as tuition) as β ,

4. Although our theoretical model considers declines in cohort quality arising from changing ability compositions, the decline in quality may also arise from sudden increases in enrollment reducing the quality of college education when resources are relatively fixed in the short run. Our empirical work can also address these types of changes in cohort quality.

FIGURE 2
Educational Attainment by Birth-Year Cohort: Females



and assume that it is positive and independent of ability.⁵

A worker of ability λ will decide to enter college if the benefit exceeds the cost, $R(\theta, r, \lambda) > \beta$. At each interest rate, r , one can define a cut-off ability level λ^* such that $R(\theta, r, \lambda^*) = \beta$. λ^* is a function of θ, r , and β . Further, $\partial\lambda^*/\partial\theta = -R_\theta/R_\lambda < 0$, $\partial\lambda^*/\partial\beta = 1/R_\lambda > 0$, and $\partial\lambda^*/\partial r = -R_r/R_\lambda > 0$. In other words, the cut-off ability declines as the price of college skills increases and rises with increases in tuition costs and the interest rate. Denoting the joint population density of interest rate and ability as $h(r, \lambda)$, the size of the college-going population (N) and their average ability ($Ec[\lambda]$) are obtained as in equation (2).

$$(2) \quad N = \int_0^{\bar{r}} \int_{\lambda^*}^1 h(r, \lambda) d\lambda dr,$$

$$Ec(\lambda) = 1/N \int_0^{\bar{r}} \int_{\lambda^*}^1 \lambda h(r, \lambda) d\lambda dr.$$

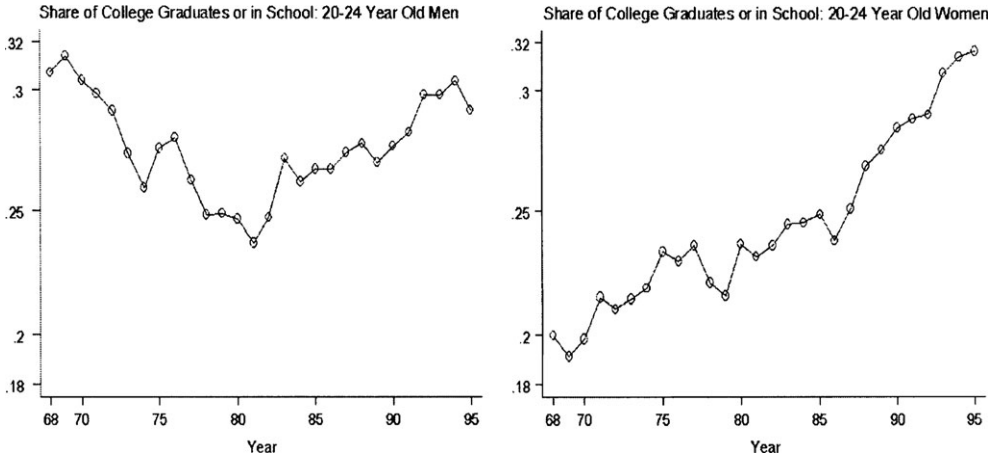
In the equation the interest rate is assumed to take a value between zero and a finite upper-bound $\bar{r} < \infty$.

*Case 1: Everybody Faces the Same Interest Rate at r^**

When everybody faces the same interest rate, the joint density can be denoted as the marginal density, $h(\lambda)$, and the number of college entrants and their average ability may be calculated by integrating only over ability, λ . Under this assumption, now consider an increase in the college skill premium, $d\theta > 0$.

5. Of course, an alternative set-up would be to assume that the return, R , does not vary with ability, but the cost of going to college is negatively related to ability.

FIGURE 3
Share of 20–24 Year-Olds in School or College Graduates



The change in the skill premium induces a change in the cut-off ability level, λ^* , and it can be shown that $\partial N/\partial\theta = -h(\lambda^*) \cdot \partial\lambda^*/\partial\theta > 0$. That is, college enrollment increases. At the same time, $\partial E_c(\lambda)/\partial\theta = (E_c[\lambda] - \lambda^*)h(\lambda^*)/N \cdot \partial\lambda^*/\partial\theta < 0$, and the average ability of college entrants declines. An increase in returns to schooling induces relatively low-ability workers to enter college. In this simple case, we will see a negative relationship between the number of college graduates and their average ability. An increase in tuition costs that does not vary with ability, $d\beta > 0$, would be the mirror image of the case described. College enrollment will decline and average ability will rise. These simple cases are illustrated in Figure 4.

Case 2: Individuals Face Different Interest Rates Due to Borrowing Constraints.

In the presence of imperfect capital markets, individuals can also differ in terms of their access to funds. It may be the case, for example, that individuals from wealthier families have lower cost of borrowing than individuals from poorer families. In this case, the cut-off ability $\lambda^*(\theta, \beta, r)$ rises with r , and it is possible that very high-ability individuals will choose not to attend college when faced with very high borrowing costs. If a given increase in the returns to college skills, θ , or reduction in tuition costs, β , leads to greater

adjustments among individuals facing higher borrowing costs, it is at least theoretically possible that these marginal entrants will actually raise the average ability of the college-educated group. In the following we characterize the conditions under which such cases can be ruled out.

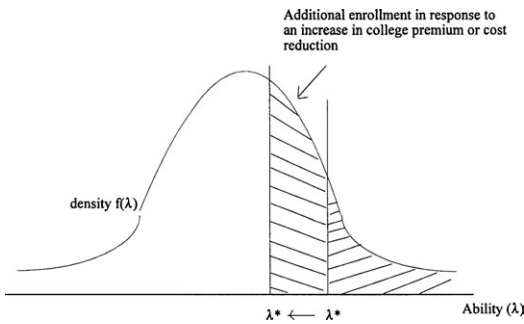
The impact of a change in the skill price on the number of college entrants and their ability are obtained by differentiating (2) with respect to θ as in the following:

$$(3) \quad \begin{aligned} \partial N/\partial\theta &= \int_0^{\bar{r}} -h(r, \lambda^*)\partial\lambda^*/\partial\theta dr > 0 \\ \partial E_c(\lambda)/\partial\theta &= 1/N \int_0^{\bar{r}} \{E_c(\lambda) - \lambda^*\} \\ &\quad \times \partial\lambda^*/\partial\theta h(r, \lambda^*)dr. \end{aligned}$$

The first partial derivative in equation (3), $\partial N/\partial\theta$, has an unambiguously positive sign because λ^* is a decreasing function of θ . An increase in the college skill premium increases the number of college entrants.

If the second partial derivative in equation (3) is negative, this would lead to negative co-variation between the size and average ability of the college educated group. Because $\partial\lambda^*/\partial\theta < 0$, the second partial derivative in equation (3) is negative when $E_c(\lambda) - \lambda^*$ is positive at all levels of r . In other words, the cut-off ability level is below the average ability of college-educated group at all levels of r . $E_c(\lambda) - \lambda^*$ is likely to be positive at all levels of r if the

FIGURE 4



cut-off ability does not vary much across interest rates, that is, $\partial\lambda^*/\partial r$ is positive but small. Because $\partial\lambda^*/\partial r = -R_r/R_\lambda$, this condition is satisfied when the comparative advantage in returns to higher ability individuals, R_λ , is large relative to the disincentive effect of the higher interest rate. In other words, ability plays a larger role in determining who acquires education, and large differences in r generate only small differences in λ^* . This, in effect, gets us close to case 1, and an increase in θ will most likely lead to a decline in average ability.

Another component is the adjustment $\partial\lambda^*/\partial\theta$ at high and low interest rates. If $|\partial\lambda^*/\partial\theta|$ decreases with r so that adjustments at high interest rates are smaller than those at lower interest rates, this reduces the likelihood of an increase in average ability.

Finally, the shape of the joint density, $h(r, \lambda)$, matters. The negative relation between size and average ability of the college group is more likely if the joint density, $h(r, \lambda)$ has little mass at high interest rates and high ability. This effectively reduces the importance of the adjustment at high ability–high interest rate margins. We have assumed so far in our discussion that the distribution of ability and borrowing costs are uncorrelated in the population. Recent papers by Cameron and Heckman (1998) and Cameron and Taber (2000) suggest that this correlation may be negative. To the extent that family income affects ability directly rather than through borrowing costs, this would strengthen the negative relationship between size and average ability.

We have so far examined how college enrollment and average ability changes in response to an increase in returns to college,

$d\theta > 0$, or its mirror image, reduction in tuition costs, $d\beta < 0$. It is also interesting to consider whether size and ability would negatively covary when borrowing costs, r , change due to government loans and subsidies. We examine this case in the appendix and focus on two different types of subsidies: a general subsidy that lowers the cost of funds for everyone and a narrower subsidy that targets those individuals who would otherwise face very high interest rates. Our conclusions are intuitive and not very surprising. A general subsidy that lowers r for everyone would lead to an increase in college enrollments and an unambiguous decline in average ability. It is possible that a targeted subsidy that only affects those who initially face very high interest rates may lead to an increase in average ability. This is due to the fact that these individuals at the margin are high-ability individuals, and inducing them to enroll in college may actually increase the average ability of the college population.⁶

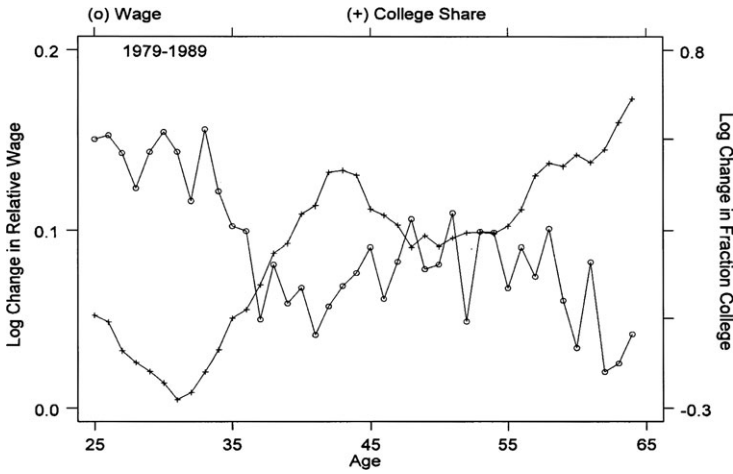
IV. THE EFFECT OF COHORT-SPECIFIC COLLEGE SHARE ON THE RELATIVE WAGE OF COLLEGE GRADUATES

In this section, we examine the relationship between rising college share and average ability of the college educated group using cohort-level data from the 1940–90 Census. Our analysis covers a long span of time using comparable data, which is an advantage. A disadvantage is that we do not have direct measures of ability and therefore must infer changes in ability through changes in wages. More specifically, we examine whether college graduate men from highly educated cohorts earn a smaller college premium, holding constant other variables that also affect the college premium.

In what follows, we first aggregated the data by census year, single year of age, and by college completion. We ran regressions in changes using weighted least squares on this aggregate-level data. Our dependent variable is the de-

6. Cameron and Heckman (1998) and Cameron and Taber (2000) conclude that credit constraints are less important than permanent income and family background characteristics in explaining schooling attainment. Both papers find that borrowing constraints are not empirically important, at least for the recent cohorts. These studies suggest that short-term tuition subsidies are unlikely to encourage enrollment of high-ability students from lower-income families.

FIGURE 5
Log Change in Relative Wage and College Share



cade change in log relative wage of college graduates at different ages. In other words, we compare the college premium received by, for example, 40-year-old college graduate men in year t with the college premium received by 40-year-old college graduate men 10 years earlier, thereby utilizing between-cohort variation in the data. Our main regressor of interest is the between cohort log differences in the share who went to college, $\Delta \log(N_{ac}/N_a)$. We use predicted college share of each cohort at age 35 as specified in equation (1) rather than actual observed shares at time t , so as to not confound the effects of cohort quality and relative supply. The variation in the data that we exploit is illustrated in Figure 5, which graphs decade log changes in relative wage and college share at different age levels over the period 1979–89. The picture illustrates the strong negative correlation between changes in relative wage and college share during this latter period. Although they are less pronounced, negative correlations also exist in other periods, as illustrated in Figure 6.

To the extent that workers in different education groups are imperfect substitutes for each other, the increase in college share will negatively impact the relative wage of college graduates regardless of the quality effect. We control for relative supply in three alternative ways. In the first specification, we include year dummies to control for aggregate supply

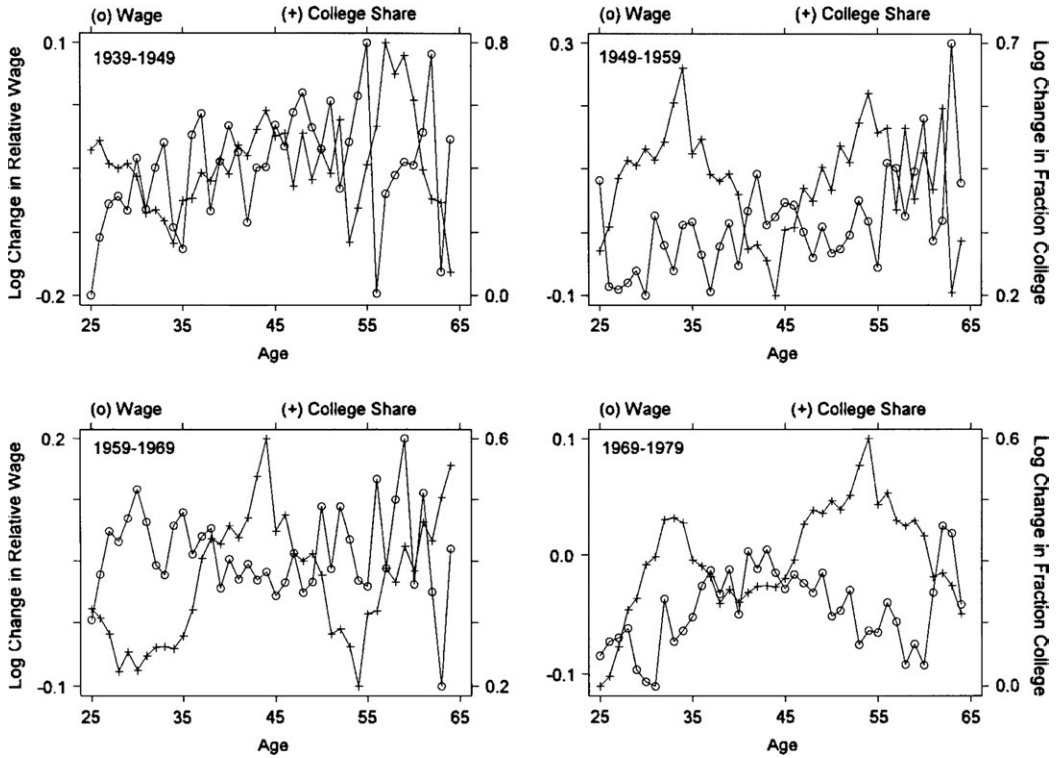
and demand shocks and run the following regression

$$(4) \quad \Delta \log(W_{act}/W_{at}) = \beta_1 \Delta \log(N_{ac}/N_a) \\ + \beta_2 \Delta \log(N_{act}/N_{ct}) \\ + \tau_t + \delta_a + \varepsilon_{at},$$

where a indexes age and c indexes college graduate. This assumes that workers in different age groups are perfect substitutes for each other so that an influx of new college graduates will have an equally adverse impact on all college graduates, and the year dummies control for this effect. As indicated in equation (4), we also include a control for overall cohort size, $\Delta \log(N_{act}/N_{ct})$, and dummies for young (25–35) and middle age (35–50).

In the second specification, we control for changes in the relative supply of workers with different education levels, high school dropouts $\Delta \log(N_{dt}/N_t)$, high school graduates, $\Delta \log(N_{ht}/N_t)$, and college graduates, $\Delta \log(N_{ct}/N_t)$. The relative supply of college graduates is the actual share observed in year t , rather than the predicted share at age 35, which is our main cohort quality variable. However, the two variables are highly collinear, making it difficult to separately identify the two effects. We follow the strategy suggested in Welch (1979), where we define relative supply as the weighted average around

FIGURE 6
Log Change in Relative Wage and College Share



a given age group. The supply of college graduates, for example, is defined as $N_{act}^* = 1/9 * (N_{a-2,ct} + 2*N_{a-1,ct} + 3*N_{act} + 2*N_{a+1,ct} + N_{a+2,ct})$. This assumes that workers in nearby age groups are closer substitutes. More specifically, we run the following regression for this specification:

$$(5) \quad \Delta \log(W_{act}/W_{at}) = \beta_1 \Delta \log(N_{ac}/N_a) + \beta_2 \Delta \log(N_{act}/N_{ct}) + \beta_3 \Delta \log(N_{adt}^*/N_{at}) + \beta_4 \Delta \log(N_{aht}^*/N_{at}) + \beta_5 \Delta \log(N_{act}^*/N_{at}) + \tau_t + \delta_a + \varepsilon_{at}$$

Finally, we include a relative supply variable based on years of potential experience rather than age. In other words, in this specification, we assume that college graduates compete with high school graduates, who are typically about four years younger and en-

ter the labor market in the same calendar year. We run the following regression in the last specification:

$$(6) \quad \Delta \log(W_{act}/W_{at}) = \beta_1 \Delta \log(N_{ac}/N_a) + \beta_2 \Delta \log(N_{act}/N_{ct}) + \beta_3 \Delta \log(N_{xdt}/N_{xt}) + \beta_4 \Delta \log(N_{xht}/N_{xt}) + \beta_5 \Delta \log(N_{xct}/N_{xt}) + \tau_t + \delta_a + \varepsilon_{at}$$

where N_{xt} refers to the number of workers at a given year of potential experience.

The regression results are reported in Table 2. Column (1), which corresponds to our first specification, shows that the coefficient on the cohort college share variable is negative and statistically significant when only year dummies are included to control for relative supply. However, the relationship between cohort college share and the relative wage of

TABLE 2
Effects of Cohort-Specific College Share on Wages of College Graduate Men
(Dependent Variable = $\Delta \log[W_{act}/W_{at}]$)

	(1)	(2)	(3)
Cohort college share	-0.101* (0.027)	-0.063 ⁺ (0.038)	-0.060 (0.039)
Relative supply			
< 12 years	—	0.056 ⁺ (0.030)	0.026 (0.030)
12–15 years	—	-0.027 (0.040)	0.031 (0.037)
16+ years	—	-0.075 ⁺ (0.040)	-0.037 (0.032)
Cohort size	-0.042* (0.019)	-0.000 (0.025)	-0.025 (0.022)
Year dummies included	yes	yes	yes
Age-group dummies included	yes	yes	yes
Relative supply (age)	no	yes	no
Relative supply (experience)	no	no	yes
Number of observations	200	200	200

Notes: Wage regressions also include control for cohort size, five year dummy variables, and two age dummy variables for young and middle age workers. Cohort college share = $\Delta \log(N_{act}/N_a)$. Cohort size = $\Delta \log(N_{act}/N_{ct})$. Relative supply in column (2) is defined as $N_{act}^* = \{1/9*(N_{a-2,et} + 2*N_{a-1,et} + 3*N_{aet} + 2*N_{a+1,et} + N_{a+2,et})\}/N_{at}$ where e refers to the education group high school dropouts, high school graduates, or college graduates. Relative supply in column (3) is N_{xet}/N_{xt} where x refers to years of potential experience rather than age. (*) denotes significance at 5% level. (+) denotes significance at 10% level. Source: 1940–1990 Census PUMS (1% sample). The sample includes 25–64 year-old men who were working during the survey week.

college graduates becomes much weaker in our alternative specifications shown in columns (2) and (3). When we use the relative supply measure defined around age (column [2]), the cohort college share variable is marginally significant (at the 10% significance level). The coefficient -0.063 implies that the average premium is about 0.6% lower for a cohort that is 10% larger than average. When we use our alternative relative supply measure, which is based on experience (column [3]), the size of the coefficient on cohort college share variable is similar (-0.060), although it is no longer significant.

A large body of literature on the effects of cohort size on earnings suggests that the wage penalty associated with being in a large cohort is especially pronounced on labor market entry as suggested by Welch (1979) and Berger (1985). In the following section we examine whether our cohort quality effect also varies by age. We ran regressions as specified in (5) separately for young workers (between the ages 25 and 34) and for older workers (35 and older). The coefficients are reported in Table 3. When we examine young workers only, our sample is reduced to 50 observations, and we find that the cohort quality variable is no longer significant. In the specification where we control for relative supply (column [2] of Table 3), the cohort quality variable ac-

tually turns positive and is marginally significant. The cohort quality variable is much more robust and stable when we look at older workers (columns [3] and [4] of Table 3). Aside from the fact that we have more observations for the older group, our assumption that workers at different ages are equally good substitutes for each other may be more valid for older workers, thus allowing us to more cleanly separate the quality and quantity effects.

What is the economic significance of a coefficient such as (-0.063)? What does this imply about the possible role of declining cohort quality in the 1970s and the 1980s? To answer this question, we use the coefficients from our cohort-level regressions (column [2]) to decompose total decade log changes in the relative wage of college graduates into components due to cohort quality, relative supply, and year effects. Overall, the relative wage of college graduates declined 5.3% during the 1970s and increased 9.3% during the 1980s. Of the 5.3% decline over the 1970s, relative supply changes contributed a decline of 6.6%, whereas an additional 2% decline was predicted by changes in the share of college. Aggregate time effects contributed an increase of 3.3% over this period. During the 1980s, the impact of cohort quality was smaller. Aggregate time effects contributed an increase of 13.6% in the relative wage of college

TABLE 3
 Effects of Cohort-Specific College Share on Wages of College Graduate Men
 (Dependent Variable = $\Delta\log[W_{act}/W_{at}]$)

	Workers Aged < 35		Workers \leq 35	
	(1)	(2)	(3)	(4)
Cohort college share	-0.039 (0.057)	0.188 ⁺ (0.098)	-0.068 ⁺ (0.036)	-0.075 (0.048)
Relative Supply				
< 12 years	—	0.222 (0.161)	—	0.014 (0.056)
12–15 years	—	0.542	—	-0.068 (0.040)
16+ years	—	-0.243 (0.271)	—	-0.043 (0.048)
Cohort size	0.025 (0.054)	0.094 (0.069)	-0.052 ⁺ (0.029)	-0.037 (0.035)
Year dummies	yes	yes	yes	yes
Age dummies	no	no	no	no
Relative supply (age)	no	yes	no	yes
Number of observations	50	50	150	150

Source: 1940–90 Census PUMS (1% sample). The sample includes 25–64-year-old men who were working during the survey week.

graduates. Relative supply changes contributed a decline of 3.3% and changes in college share contributed a decline of 1%. We conclude from Table 2 that the cohort college share variable (our measure of cohort quality) is only weakly related to relative wages of college graduates once we control for relative supply. Overall, relative supply and time effects, which may reflect aggregate demand changes in favor of college graduates, account for most of the variation (87%) in college wages.⁷

It may be useful to compare our results based on aggregate data to microlevel evidence from test score results. Recently, Murnane et al. (1995) reported test results from two successive cohorts using the National Longitudinal Survey 1972 and the High School and Beyond. They report average item-response theory-scaled math scores for college graduates and for the entire sample who finished high school. They find that between those two cohorts the college versus overall test score differential increased 1.39 points, implying an increase in the average

ability of college graduates between these two cohorts. Because the share of the cohort who attended college fell during this period, this is consistent with a negative relationship between size and average ability. According to their estimates, a 1-point increase in the math score results in a 0.013 increase in log hourly wages six years after the test. This results in a small predicted increase in the relative wage of college graduates between these two cohorts, generally in line with the small quality effects we find herein.

Finally, we end with the question we began with. What do we think of the enormous increase in the share of college graduates over time? Despite the fact that the cohort quality effect we estimate is small, nevertheless this would imply a large impact of declining cohort quality on college wage premium over 1940–90 as the share of college graduates increased from 5% in 1940 to 23% in 1990. We believe that this exercise would overstate the importance of cohort quality effects because the key assumption is that the distribution of ability is fixed over time. If we adopt a more broad definition of ability, as accumulated knowledge of potential college entrants, it seems very reasonable that ability would respond to changes in the demand for college graduates over the long run. Thus, our theoretical and empirical work may be more appropriate for examining short-run changes where the distribution of ability can be described as fixed.

7. Recently, a number of papers have examined whether the recent increase in the college premium was due to rising returns to raw ability rather than skills acquired in college. See, for example, Blackburn and Neumark (1993), Chay and Lee (1997), Cawley et al. (1998). Although this is an important question in its own right, our focus in this article is on wage effects due to changes in ability *composition* rather than changing returns. In our empirical work, both increases in returns to ability and college skills are captured by the aggregate time effects reported.

V. SUMMARY

Using Census data, this article documents the enormous increase in educational attainment of men and women in the United States during the 1940–90 period. The article examines whether the increase in number of college graduates lead to a decline in the average quality of college graduates and consequently affected their relative wages. We find some weak evidence that college graduates from more educated cohorts receive a smaller college premium, holding constant other factors that also affect relative wages.⁸ These quality effects, however, appear to be small relative to the effects of demand and supply shocks and appear to have played only a minor in the recent fluctuations in the college premium.

APPENDIX

In this section, we consider changes in the distribution of borrowing costs, r . For simplicity of presentation, we assume that the distributions of r and λ are orthogonal. Under that assumption, the joint density $h(r, \lambda)$ can be written as a product of marginal densities, $g(r|\delta)f(\lambda)$. Here δ is the shift parameter specific to the density of interest rate, $g(\cdot)$. A change in δ induces changes in the number of college entrants and their average ability as the following:

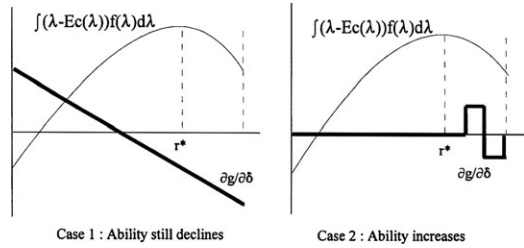
$$\begin{aligned}
 (A-1) \quad \partial N / \partial \delta &= \int_0^{\bar{r}} \left[\int_{\lambda^*}^1 f(\lambda) d\lambda \right] \partial g / \partial \delta dr \\
 &= \int_0^{\bar{r}} N(r) \partial g / \partial \delta dr \\
 \partial E_c(\lambda) / \partial \delta &= 1/N \int_0^{\bar{r}} \left[\int_{\lambda^*}^1 (\lambda - E_c[\lambda]) f(\lambda) d\lambda \right] \\
 &\quad \times \partial g / \partial \delta dr.
 \end{aligned}$$

In this expression, $N(r)$ is the number of college entrants facing the interest rate r and decreases with r . The impact on the number of college entrants depends on the correlation between $N(r)$ and $\partial g / \partial \delta$. Relaxation of borrowing constraints will tend to make $\partial g / \partial \delta$ positive at low r and negative at high r , and thus the number of college entrants will increase given such change in the distribution of r .

The impact on average ability depends on the correlation between $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda$ and $\partial g / \partial \delta$. As $E_c(\lambda)$ is the weighted average of $\int_{\lambda^*} \lambda f(\lambda) d\lambda$ where the weights are $g(r|\delta)$, the integral must have a negative value at low r and a positive value at high r . As a result, the integral must increase with r initially. But at high r , it may start to de-

8. In earlier versions, we also examined the occupation distribution of college graduates from different birth cohorts. We found some evidence, particularly in the more recent data, that college graduates from more educated cohorts were less likely to be working in the most skilled occupations, where occupations are ranked based on the average wage in the occupation. These results are available on request from the authors.

FIGURE A1



cline as λ^* may exceed $E_c(\lambda)$. Consequently, $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda$ has an inverted U shape and the impact on average ability may vary with the extent of relaxation in borrowing constraints.

A general relaxation in borrowing constraints may be parameterized as positive values of $\partial g / \partial \delta$ at low r and negative values of $\partial g / \partial \delta$ at high r . This is described in the left panel of Figure A-1. In this case, $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda \cdot \partial g / \partial \delta$ tends to have a negative value, and as a result, the average ability tends to decline.

When the relaxation is of a very limited scope, the average ability may increase. This case is described in the right panel of Figure A-1. Suppose the borrowing constraints are relaxed only for those with extremely high interest rates. Then $\partial g / \partial \delta$ will be negative at such high interest rates, and it will be positive at slightly lower interest rates. If this change in the density $g(r|\delta)$ occurs where $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda$ decreases, a higher value of $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda$ is multiplied with positive $\partial g / \partial \delta$ and a lower value of $\int_{\lambda^*} (\lambda - E_c[\lambda]) f(\lambda) d\lambda$ is multiplied with negative $\partial g / \partial \delta$. Consequently, the average ability may increase.

The intuition behind these results is quite simple. The change in average ability depends on those newly entering college. In the first case, the change in density is assumed to induce more college enrollment among those who have faced relatively low interest rates. At the margin those individuals at low interest rates are less able individuals and their entry decreases average ability. In the second case, the change in density induces actions only at high interest rates.

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