Finish It and It’s Free: An Evaluation of College Graduation Subsidies

Matthew D. Webb*

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Abstract

Despite the rapid increase in the returns to higher education witnessed in the labor market over the past few decades, there has also been a marked increase in the share of individuals who dropout of college or university. To boost student persistence in higher education, several Canadian provincial governments introduced a set of reforms that were designed as subsidies for college graduation. In addition, these policies were designed to discourage internal migration following graduation. Using data from both administrative tax records as well as longitudinal surveys, I analyze the effectiveness of these policies. The main findings are that the programs had no effect on internal migration, but significantly reduced college dropout rates.

Keywords: higher education, college dropout, some college, college attrition, education financing

JEL Codes: I22, I23, I28

* Department of Economics, Carleton University, Ottawa, Ontario, Canada, K1S 5B6. Email: matt.webb@carleton.ca. I am thankful to Michele Campolieti, Marco Cozzi, and Allan Gregory, Steven Lehrer, and James MacKinnon for thoughtful evaluations on a prior draft. I am grateful to participants at the 40th and 42nd ACEA conferences, the 47th CEA meetings, the 84th SEA Conference, the 21st SOLE conference, the 2015 World Congress of the Econometric Society, and the 2015 Canadian Public Economics Group meeting. I appreciate the suggestions from Haggay Etkes, Julian Hsu, Mohsen Javdani, Joniada Milla, Emilia Simeono, and Fraser Summerfield. Thank you to David R. Johnson and Rene Morissette for providing me with data and code. Part of this analysis was conducted at the Prairie Regional RDC which is part of the Canadian Research Data Centre Network (CRDCN), I am thankful for the assistance of Bin Hu and Rebecca Williams. Paul Roberts’ help at Statistics Canada is also greatly appreciated. I gratefully acknowledge financial support from the Social Sciences and Humanities Research Council of Canada.
1 Introduction

In the past several decades there has been a large increase in college and university enrollment. There has also been a large increase in the college dropout rate (Turner, 2004). Furthermore, completion rates have stagnated among recent cohorts, as discussed in Oreopoulous and Petronijevic (2013). Currently, the six-year graduation rate in four-year colleges in the US is 58% and the seven-year completion rate at a large sample of Canadian universities is 84%.\(^1\) With dropout rates being as large as they are, many have started to ask what can be done to increase college persistence.\(^2\) This paper examines a set of tax policies which included strong incentives for college graduation, offering tax credits totalling between $15,000 and $25,000. Four Canadian provinces have offered these credits to recent graduates in an effort to curb outflows of graduates to other provinces. This paper analyzes the impact that these programs had on college enrollment, college dropout, and migration decisions.

These programs have similar aims to the various Merit Scholarship programs offered by several American state governments.\(^3\) Both the Merit Scholarships and the retention credits attempt to raise the average level of educational attainment within a jurisdiction. The merit programs and the retention programs both tie educational funding to a geographic location but the merit programs do not restrict where recipients can reside post-graduation. The Merit Scholarships offer incentives to enroll in college in a specified location, while the retention credits offer incentives to graduate from college and to reside in a specified location after graduation. Additionally, the Merit Scholarships offer financial assistance during the years of college enrollment, while the retention credits offer assistance after graduation. While the Merit Scholarships have received considerable attention, there has been very little analysis of the Graduate Retention Credits.\(^4\) This paper provides the first analysis of causal outcomes of these programs.

Concern has been given to the local stock of human capital in recent years, as an important factor for regional economic growth and development. In particular,

\(^1\)The American figure is from National Center for Education Statistics (2012) and the Canadian figure is the author’s calculation using data from Maclean’s Magazine (2014).


\(^3\)Fifteen states now offer some form of merit scholarship, and the effectiveness of these various programs has been studied by Fitzpatrick and Jones (2012).

\(^4\)Essaji and Neill (2010) detail the costs and program features of these credits but not the consequences of these programs.
several researchers have studied the effectiveness of various policies at increasing the local stock. The lock stock of graduates has proved difficult to increase as higher levels of education also increase an individuals’ likelihood to emigrate within country (Malamud and Wozniak, 2012). Other research has suggested that Merit Scholarships do not increase the retention of college-bound students after graduation (Sjoquist and Winters, 2013b). In part, the retention programs were introduced to discourage internal migration since historically many graduates have moved provinces after they graduate. This paper additionally analyzes whether the programs changed migration patterns of recent graduates, and finds the patterns unchanged.

Using both administrative and survey data the programs are found not to have an impact on migration decisions. This perhaps explains why one of the provinces, Nova Scotia, recently cancelled its graduate retention program. This paper also introduces a new proxy for college graduation into the Longitudinal Administrative Databank. The analysis shows that the programs did have a significant impact on reducing college dropout rates, by 4.1% among individuals 18-23. Interestingly, there was no estimated increase in the likelihood of enrolling in college. These results are opposite of what is typically found when tuition is reduced, where enrollment increases, but completion rates stagnate.

The remainder of this paper is as follows, Section 2 places the programs in the context of the literature. Section 3 gives a detailed overview of the various graduate retention programs under study. Section 4 describes the linear Difference-in-Differences (DiD) estimation strategy and the relevant treatment and comparison groups. As inference with DiD is difficult, especially with few treatment and comparison groups, the methodology for inference is discussed in section 4.3. Section 5 describes the administrative and survey data sources and the outcomes of interest. Results are discussed in Section 6, which presents evidence from both data sources that the programs had a significant impact on decreasing college dropout rates, but did not decrease out of province migration. Section 7 concludes.

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6 The Saskatchewan government mentions that there “has been significant leakage of post-secondary graduates outside the province” (The Saskatchewan Labour Market Commission’s, 2009).
8 Cohodes and Goodman (2014) find completion rates decreased among those treated by the Massachusetts Merit Scholarship.
2 Context of the Retention Programs

This paper investigates how individuals respond to subsidies for college graduation. Starting in 2005, several Canadian provinces offered ‘Graduate Retention Credits’, in an effort to discourage recent graduates from moving out of province. The design of these programs offered large tax credits which were conditional on graduation. The programs are explained in greater detail in Section 3. In general, the programs offered between $15,000 and $25,000 in tax credits to recent university graduates. The credits refunded between 50% and 97% of average four-year total tuition in these provinces. This paper examines the impact that these programs had on a host of education and migration decisions. It finds that the programs decreased dropout rates by 4.1%.

The witnessed increase in dropout rates is surprising given that the relative returns to education have increased over the same time period. Deming and Dynarski (2009) find that real wages for high school graduates fell by one-third from 1972 to 2005, while real wages for college graduates held steady. For men, the bachelor’s degree wage premium was 22% in 1972 and it had increased to 60% by 2003. Findings such as these, lead Oreopoulou and Petronijevic (2013) to summarize that although the returns are heterogeneous, college is a worthwhile investment for both the average and the marginal student.

The increase in the dropout rate would not be as worrying if the benefits of going to college increased linearly in years of education. However, it has been known since Hungerford and Solon (1987) that there are “degree effects” or “sheepskin effects” conferred on those who graduate. More recent literature has found that degree effects exist across the distribution of earnings, Oreopoulou and Petronijevic (2013), and that earnings for those with some-college are only slightly higher than the earnings of high school graduates. Within Canada, Ferrer and Riddell (2008) estimate sheepskin effects in wages for holders of bachelor’s degrees on the order of 20% for women and 16% for men. Additionally, Riddell and Song (2011) find that re-employment prospects, following a job termination, are higher for college graduates. Finally, Jepsen, Troske and Coomes (2014) find both earnings premiums and higher levels of employment for community college graduates. The prevalence of sheepskin ef-

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9The literature on sheepskin effects is vast and has found among other things that the effects are larger for females and minorities, Belman and Heywood (1991), Ferrer and Riddell (2008); and exist in several countries, Denny and Harmon (2001).
ffects across a variety of countries and education levels suggests that individuals who
dropout possibly do so at significant private cost.\textsuperscript{10} For example, Webber (2016) es-
timates that for the median student the net present value of a degree to be between
$85,000 - $300,000 depending on the major. That the programs decreased dropout
rates is surprising, given the large rewards in the labor market for graduating.

In addition to the private costs there are also social costs from dropping out. Schneider and Yin (2012) perform a ‘back of the envelope’ calculation and suggest
large losses in income tax revenue from dropouts. There is also speculation that many
new jobs will require higher levels of education suggesting diminished employment
prospects for dropouts going forward. According to the United States Bureau of
Labor Statistics, job growth in occupations requiring some post secondary education
is expected to outpace job growth in occupations requiring a high school education
or less over the coming decade (Bureau of Labour Statistics, U.S. Department of
Labor., 2012).

This paper investigates whether the programs were able to retain recent gradu-
ates, as the programs were intended to do. There are many reasons why a provincial
government would want to retain these individuals. The post-secondary education
(PSE) spillovers literature presents evidence that higher levels of education benefits
the community at large. When analyzing minimum schooling laws, Acemoglu and
Angrist (2001) find evidence that there are small positive external returns to an extra
year of education. Moretti (2004) uses the presence of a land grant university as an
instrumental variable for college attendance to examine the impact of increasing the
share of college graduates in a city. He shows that a percentage point increase in the
presence of college graduates is associated with increased wages for others: specif-
ically a 1.9\% wage increase for high-school dropouts and a 1.6\% increase for high
school graduates. In a similar study, Shapiro (2006) examined American data at the
metropolitan level and finds that a 10\% increase in a city’s concentration of college
graduates was on average followed by a 0.8\% increase in employment growth. Aghion
et al. (2009) find that investment in four-year college educations has a positive effect
on economic growth.

The programs are unique in that they place no restrictions on pre-college res-
idency, or on where one studies, but restrict where one works after graduation.

\textsuperscript{10}However, it is possible that the dropout decision is welfare improving as simulations in Stine-
brickner and Stinebrickner (2013) show that newly enrolled students who perform poorly learn that
staying in school is not worthwhile.
While most education funding for students has been offered with few geographical constraints, there have been a few exceptions. There are the aforementioned Merit Scholarships, and differential tuition levels for in-state and out-of-state students. Some jurisdictions, including Canadian provinces, have experimented with targeted retention/attraction programs to attract individuals in certain occupations, such as doctors and nurses.\textsuperscript{11} Maryland has offered scholarships to residents which require an individual to work one year in the state for each year they receive a scholarship (Groen, 2011). In 2012, Kansas offered incentives such as student debt repayments and income tax waivers to attract individuals to rural Kansas. Unlike the graduate retention programs, to be eligible for the incentives individuals must prove they have resided outside of the state for at least the previous five years.\textsuperscript{12} Finally, in 2007, Maine introduced a program generally similar to the retention credits studied here. This program repaid up to $5500 per year in student loans for bachelor degree holders from a Maine college.\textsuperscript{13} To my knowledge, the outcomes of either the program in Maine or Kansas have yet to been analyzed.

Given the large value of these credits, it is interesting to compare the programs to other large scale education financing reforms. Gunnes, Kirkebøen and Rønning (2013) investigate an experiment in Norway where students were offered a financial incentive of $3000 (USD) if they graduated ‘on time’. They find that the incentive reduced mean graduation delay by 0.23 semesters per year treated. Similarly, Garibaldi et al. (2012) find that at Bocconi University a 1000 € increase in tuition reduced the likelihood of late graduation by 5.2%, with no increase in the dropout rate. Arendt (2013) finds that a large increase in grants decreased dropout rates, but had no impact on completion rates after controlling for various student and parental characteristics. Dowd (2004) finds that an increased amount of subsidized loans in the first year of college enrollment increased persistence. Recently, Denning (2017) found that additional financial aid accelerates graduation rates for university seniors.

Past literature on financial aid suggests that it can be difficult to predict the impact of a reform as administrative details and program knowledge are important. Deming and Dynarski (2009) show that ‘financial aid’ is not a homogeneous good and that paperwork matters. Programs with high administrative hurdles have smaller

\textsuperscript{11}See Reamy (1994) and Mullan (1999).
\textsuperscript{12}For more information of the Kansas program see http://www.kansascommerce.com/index.aspx?nid=320
\textsuperscript{13}For more information see http://www.opportunitymaine.org/opportunity-maine-program/frequently-asked-questions/.
benefits. The retention credits are relatively easy to apply for, and in some provinces can be claimed as part of provincial tax returns. The salience of these programs, especially among high school students, is an open question. McGuigan, McNally and Wyness (2012) surveyed high school students in London, England after recent tuition reforms and found that roughly half of the students surveyed were unaware of key features of the reforms.

3 Background of the Graduate Retention Programs

The four Canadian provinces that have implemented graduate retention programs are Saskatchewan, Manitoba, New Brunswick and Nova Scotia. Table 1 provides an overview of the various program attributes and how they differ across provinces. Broadly speaking the programs are quite similar, and the amounts they offer to college and university graduates are of the same order of magnitude. All of the programs are income tax credits, though the characteristics of the credits differ across provinces. Only one of the credits is refundable, though most of the credits do roll-over.

Saskatchewan and Manitoba do not require a separate application for the graduate retention credits, which can be claimed on income tax returns. Nova Scotia and New Brunswick require separate applications to claim the credits. Three of the four programs determine the maximum credit based on the amount of tuition paid, while Nova Scotia offers a fixed amount to each recent graduate. The proportion of tuition refunded in each province varies, with up to 100% of tuition being refundable in Saskatchewan and 50% being refundable in New Brunswick. The maximum amount of the credits is the same in both of these provinces, but given the differing tuition refund percentages a student would have had to pay $40,000 in tuition to receive the maximum credit in New Brunswick, but only $20,000 in Saskatchewan. Finally, the

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14 See Essaji and Neill, 2010 for a summary of the various GRP programs.
15 Quebec also operates a smaller wage subsidy program for people in remote, resource-rich regions who work in the resource industry, with eligibility contingent on holding a degree related to your current occupation. Given the specificity of this program it is ignored in the analysis. For more details about the Quebec program see http://www.revenuquebec.ca/en/citoyen/credits/credits/credits_reduisant/nouv_diplome/
16 Specifically until 2012 Saskatchewan offered a refundable credit, which meant that if individuals earned insufficient income to claim the maximum annual amount, they would receive the difference in the form of a refund. In 2012, Saskatchewan allowed the credit to roll-over, but ended the refundability provision. All but Nova Scotia offers a roll-over provision, which allows the unused portion of the credit to accrue over time. This means that eventually an individual will be able to claim the maximum allowable credit.
total costs of each program are broadly similar in each of the provinces, ranging from $24-$35 million per year. These figures are estimated to increase over the coming years as many of these programs are new and there are not yet six or seven cohorts of graduates claiming these credits. In early 2014, the government of Nova Scotia eliminated its Graduate Retention Rebate. The government cites the failure of the programs to retain graduates as the reason for cancelling the programs.  

This is perhaps not surprising as results in Section 6 suggest that the various programs did not alter internal migration decisions.

Aside from the retention credits the provincial and federal governments invest heavily in both post-secondary institutes and students. Essaji and Neill (2010) provide a summary of the characteristics and costs of the various student funding programs currently in operation in Canada, and thoroughly summarize the graduate retention programs. In addition to the programs mentioned in that paper, several provinces have recently adjusted their approach to funding post-secondary education. For instance, Ontario previously introduced a tuition credit where students are eligible for a 30% rebate of their Ontario college or university tuition. More recently, this grant was replaced by an across the board 10% reduction in tuitions. Additionally, proposed tuition increases in Quebec resulted in student strikes lasting for over six months. Further information about the details and costs of government funding for post secondary education can be found in Neill (2013).

The graduate retention programs are operationally quite different than the other means of government funding for post-secondary education. The most distinct feature is that they base eligibility on graduation. Another unique feature is that their benefits are provided solely after graduation. Ontario’s rebates refunded tuition in the year that it was paid. Similarly, the federal and provincial tuition tax credits and education amounts are claimed during the years enrolled in school. Both of

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18 The grant tops out at $1680 for university tuition and $770 for college tuition, provided that their parents earn $160,000 or less per year, see https://osap.gov.on.ca/OSAPPortal/en/PostsecondaryEducation/Tuition/index.htm for more details.
20 A recent summary of American tax benefits for college attendance can be found in Dynarski and Scott-Clayton (2016). Bergman, Denning and Manoli (2019) in a randomized control trial finds that emailing students about potential tax breaks does not affect enrollment decisions.
21 Similar programs in the United States were recently analyzed in Hoxby and Bulman (2016) and were not found to increase college enrollment.
these are non-refundable so they typically do not ease a student’s budget constraint until after graduation, as explained in Neill (2013). Student’s budget constraints are a current concern in education policy, since it has been shown that increased financial aid improves college enrollment (Sartarelli, 2011). The retention credits do nothing to ease this constraint while enrolled, but expand a student’s post graduation budget. Offsetting the less than ideal timing of the benefits, is the fact that most student loans require repayment following graduation and the payouts from the programs coincide with the repayment schedule for student loans.\textsuperscript{22}

The rebate schedules of these programs introduces an added complexity to calculating the costs of attending university. For example, if a student in Saskatchewan paid the average university tuition for four years starting in the fall of 2007, then her annual tuition would have been $5015, $5064, $5173, and $5431. Of the $20,683 paid in tuition, they would be eligible for a $20,000 rebate (or 96.69%), paid out in the following annual instalments: $2000,$2000,$2000,$2000,$4000,$4000,$4000. Determining the net present cost of such a decision is perhaps not the easiest task for a student in their final year of high school. Moreover, in most cases individuals need to have a sizeable income to receive the full amount of the credit in the least amount of time.\textsuperscript{23} Aspects of these programs are similar to the UK tuition reforms where the net present liability of tuition depends on labour market outcomes after leaving school.\textsuperscript{24}

Another distinct feature of these programs is that they are place dependent. This is an unusual characteristic of education funding, but it does align the subsidy with the provinces’ goals of having a well-educated labour force, rather than a large number of college and university students. While the programs are designed to retain an individual in a province, they are not targeted towards those on the margin of emigrating. As a result, the programs offer generous tuition rebates to all those who were never contemplating leaving the province. However, these rebates are conditional on graduation and thus may influence college graduation and dropout decisions. Having explained how these programs operate institutionally, it is worth considering how they may influence an individual’s decisions theoretically.

\textsuperscript{22} The analysis discussed in Section 6 directly estimates the impact of the credits on interest paid for student loans.

\textsuperscript{23} The roll-over provisions offered by many provinces mean that the entire value of the credit will eventually be received, but over a longer time horizon.

\textsuperscript{24} In the UK, the full value of tuition is loaned to students while in school, and repayment is conditional on post schooling wages. (McGuigan, McNally and Wyness, 2012)
4 Identification and Inference Strategy

I propose a simple model of how the introduction of graduate retention credits could impact an individual's decision making. I then describe how the impacts proposed in the model will be estimated using a linear difference-in-differences (DiD) equation. Finally, as inference with DiD can be difficult, the procedure for inference is discussed.

4.1 A Simple Model

Let us consider the situation of a student who has just finished high school. The individual wants to maximize her utility which is a function of both her discretionary income $\tilde{C}$ and quality of life (QoL). She is able to do so by choosing only two things: her level of education and the city in which she resides. Cities influence her quality of life, her earnings, and her taxes. Conceptually, we can imagine this individual solving the following problem:

$$
\max_{\text{educ1} \in \{hs,c,u\}, \text{educ2} \in \{hs,c,C,u,U\}, \text{city1,2,3} \in X} U_1(\tilde{C}_1, QoL_1) + \beta U_2(\tilde{C}_2, QoL_2) + + \beta^2 U_3(\tilde{C}_3, QoL_3),
$$

such that

$$
QoL_i = f(city_i),
$$

$$
\tilde{C}_1 = Y_1(\text{educ}_1, city_1) + SL_1(\text{educ}_1) - taxes(city_1) - rent(city_1) - tuition(\text{educ}_1),
$$

$$
\tilde{C}_2 = Y_2(\text{educ}_2, city_2) + SL_2(\text{educ}_2) - (1 + r) SL_1(\text{educ}_1 | \text{educ}_2) - taxes(city_2) - rent(city_2) - tuition(\text{educ}_2),
$$

$$
\tilde{C}_3 = Y_3(\text{educ}_3, city_3) - (1 + r) * SL_2(\text{educ}_2) - taxes(city_3, \text{educ}_2) - rent(city_3).
$$

The problem is to maximize her utility across three periods; two periods in which she can possibly further her education and the final period. Utility in the second and third periods is discounted by an exogenously given discount factor, $\beta$. Simplifying, assume that she can only make the following educational choices: in the first period she can choose no further education and remain a high school graduate (hs), she can attend a community college (c), or she can attend university (u). Additionally,
she can choose the city in which she will either work or attend school from the set of cities, \( X \). In the second period, she can choose to either remain in school, or to dropout of school. If she chooses not to continue her education in the first period, by assumption, she remains at education level hs. If she attended college in the first period she can either enter the workforce with some college, c, or finish her program and graduate, C. Similarly, if she attended university in the first period she can either enter the workforce with some university, u, or finish her program and graduate, U.

In the first two periods the educational decision will affect consumption in several ways. Obtaining additional education requires a good deal of time, and as such will negatively impact earnings, \( Y \), in the first period. Additional education has a direct tuition cost, which must be paid in the current period. Student loans (SL) are available to help pay tuition and provide additional consumption in the first two periods if additional education is chosen, but they must be repaid with interest, \( r \), in the third period. The individual is free to relocate to a new city after entering the workforce. Education is assumed to increase earnings in the third period. The graduate retention programs introduce an education argument to the tax function in the third period, as in certain cities the individual will face lower taxes, given earnings, than a similar person with a lower level of education on account of receiving a graduate retention tax credit. Rent is included in the model to reflect the fact that there are some non-traded goods that people must consume, the price of which is determined by the city of residence.

Reflecting on the impact of the graduate retention programs through the lens of this model reveals several dimensions in which the programs might have an impact. The most direct impact is that consumption in the third period is higher in a city where an individual is eligible for a credit, with all else being equal. Accordingly, if wages, taxes, and quality of life are all equal between two cities, then if one city offered a credit and the other did not we should expect the city with the credit to be the chosen residence. Moreover, the availability of a credit will increase overall consumption in the three periods for those who go to school, so we should expect that the introduction of a retention credit will increase educational attainment. This follows from the fact that when a location offers a credit, the credit increases overall three-period consumption for graduates. This increase in consumption makes obtaining further education relatively more attractive. A third possible impact may occur because the decrease in taxes from a credit coincides with student loan repayments. If there is heterogeneity in the amount of student loans outstanding, then its
probable that those with larger amounts of debt would be more likely to locate in a city offering a credit. Similarly, those with student loans may be able to accelerate their debt repayment if they receive a credit.

The graduate retention programs have one additional channel of influence. If people have already decided to enroll in school in expectation of receiving a credit upon graduation then the costs of dropping out of school are more severe than they would have been otherwise. In the absence of a credit the potential costs of dropping out is the forgone higher wages that one might earn after graduating. The presence of a credit increases the cost of dropping out by disqualifying an individual from receiving the credit. One can alternatively think about the credits as a fee/rebate program, wherein payment is made up front, and a rebate is offered on graduation. If one expected to receive in the third period a refund proportional to her first period tuition in period three, then the decision to dropout in period two would retroactively increase the ‘expected’ cost of going to school in period one.

Any influence that a program may have will depend on the age at which an individual was when the program was announced. The model is written as though the program is in place in period zero. This will be the case for individuals who reach the end of high school in or after the year a program is announced. However, there are individuals for which the programs were announced when they were in period one or period two. For these individuals, the enrollment decision will have already been made, thus the programs are more likely to impact the migration and graduation decisions. The empirical analysis will account for these differences. The model above is framed in terms of an individual’s decision, which matches well with the micro level data used for analysis. With these possible outcomes in mind, I now discuss the estimation strategy.

4.2 Estimating Equation

A linear difference-in-differences (DiD) approach will be used. Given that the impacts of the programs may depend on the age an individual was when the programs were announced, the analysis will be split into two parts. The first part will estimate the impact that the programs had on period one and two decisions, using data on individuals aged 18-23. The second part will estimate the impact the programs had on period three decisions, using data on individuals 23-28. Individuals who turned either 18 (or 23) in a province offering a retention credit will be regarded as ‘treated’.

In the analysis, individuals in the provinces offering graduate retention programs
listed in Table 1 will be regarded as ‘treated’, while individuals in the other provinces will be regarded as the ‘comparison’ group. The linear DiD will then compare the changes in outcomes over time in provinces with retention credits to changes in outcomes over time in provinces without. Identification using DiD requires there to be common support between the treatment and comparison groups, and common trends prior to the introduction of the programs in both groups, see DiNardo and Lee (2011) for additional details.

The sample used for analysis is different for the early period and late period analysis, given differing concerns about common support. The early period analysis, which focuses on educational outcomes, uses individuals from all Canadian provinces. The late period analysis, which focuses on migration decisions, will look at the impact of the programs in the Atlantic provinces. The Atlantic provinces offer a useful setting for conventional DiD analysis of migration. The Atlantic region is comprised of two provinces with retention programs: Nova Scotia and New Brunswick; and two provinces without: Prince Edward Island, and Newfoundland and Labrador. There is the additional benefit of interprovincial migration being nearly symmetric within the Atlantic Region. Accordingly, the assumption of common trends in interprovincial migration is more realistic when restricting the sample to the Atlantic provinces.

The impacts of the credits on early period outcomes are estimated by the following equation:

\[ Y_{ist} = c + \beta_{GRP} I[GRP \text{ Prov at 18}_{ist} \times \text{age GRP announced}_{ist}] + \text{TRENDS}_{st} + \text{PROV}_i + \text{AGE}_i + \text{BIRTH YEAR}_i + \epsilon_{ist} \]

In the equation there is a set of province dummy variables, a set of birth year dummy variables, a set of province specific time trends, and a set of age dummy variables. Aside from age dummies there are no control variables, as the unconditional impact of the programs is of primary concern. The coefficient of interest is \( \beta_{GRP} \) the DiD coefficient. This coefficient will capture the marginal impact of being in a province with a retention credit, for those who were young enough to be eligible for a credit. The variable for this coefficient is an indicator variable for the interaction

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25 As mentioned above, the majority of Canadians attend university within their own province. Accordingly, I assume that changes in the net cost of attending university in one province are unlikely to affect the decision of someone in another province to attend university.

26 See Statistics Canada, CANSIM table 051-0019.
between living in a province offering a credit and being young when the credit was announced. This variable will be equal to one for those who turned 18 in Nova Scotia in 2006 – 2013, in New Brunswick in 2005 – 2013, and in Saskatchewan or Manitoba in 2007 – 2013. The variable is set to zero otherwise. The set up for the late period estimating equation is nearly identical, but treatment status is defined by when an individual turned 23, and uses only individuals from the Atlantic provinces. This equation estimates a common treatment effect for the provinces offering graduate retention credits.

4.3 Inference Strategy

Inference with linear difference-in-differences can be a challenge. The prevalence of type I errors has been known since Donald and Lang (2007) and Bertrand, Duflo and Mullainathan (2004). The source of these problems is largely due to serial correlation in the error terms. These problems can be corrected for by clustering standard errors at the level of the policy change. Accordingly, the standard errors are calculated using a cluster robust variance estimator, clustered by province. However, the cluster robust variance estimator is unreliable when there are few clusters, or when clusters are unbalanced. In this analysis, clusters are both severely unbalanced and there are only four or ten clusters in the dataset.

To deal with the small cluster problem, Cameron, Gelbach and Miller (2008) (CGM) propose a wild cluster bootstrap which they claim works with as few as five clusters. However, Webb (2014) shows that with few clusters, the 2-point wild cluster bootstrap is unreliable. With few clusters, the CGM procedure calculates a point estimate for the $P$-value although the $P$-value is actually interval identified. Webb (2014) proposes the following 6-point distribution for use with few clusters:

$$v_g = -\sqrt{\frac{3}{2}}, -\sqrt{\frac{2}{2}}, -\sqrt{\frac{1}{2}}, \sqrt{\frac{1}{2}}, \sqrt{\frac{2}{2}}, \sqrt{\frac{3}{2}} \quad w.p. \quad \frac{1}{6}.$$  

To deal with unbalanced clusters, MacKinnon and Webb (2017) show that the wild cluster bootstrap works well. However, the simulations in that paper have a com-
paratively large number of clusters. Additionally, the simulations in Webb (2014) involve only balanced clusters. This leaves open the question of whether the 6-point distribution works well with few clusters.

MacKinnon and Webb (2017) also compare the wild cluster bootstrap to the CRVE procedure assuming the t-statistics follow a $t(G^* - 1)$ distribution. With a larger number of clusters in that paper this CRVE inference procedure understates the $P$-values relative to the wild cluster bootstrap, yet it remains to be tested how it performs with a relatively small number of clusters. To test this I compare the usual CRVE assuming the t-statistics follow a $t(G - 1)$ with the CRVE assuming a $t(G^* - 1)$ distribution, where $G^*$ is the effective number of clusters when testing the treatment model calculated using the clusteff Stata procedure described in Lee and Steigerwald (2018), based on the procedure in Carter, Schnepel and Steigerwald (2017).

I perform a small Monte Carlo exercise to test whether all these procedures work well with few clusters. The Monte Carlo exercise uses the homoskedasticity set up from Cameron, Gelbach and Miller (2008). Across 25,000 replications, data are generated by the following equation:

$$y_{ig} = x_{ig} + x_g + \epsilon_{ig} + \epsilon_g$$

Where $x_{ig}, x_g, \epsilon_{ig}, \epsilon_g$ are all drawn from a $N(0, 1)$ distribution. This set up imposes within cluster correlation in both the x-variable ($x_{ig} + x_g$) and in $\epsilon$ ($\epsilon_{ig} + \epsilon_g$). Two cases are considered, one to match the data in the early decision analysis, and one to match the late decision analysis. In both cases there are 1000 observations per replication. In the early decision analysis the data are divided among nine or ten clusters, proportional to the populations of Canadian provinces. In the late decision analysis the observations are distributed proportionally to the populations of the four Atlantic provinces. The clusters which are defined as treated thus have proportional size to treated provinces in the late decision and early decision analysis.

Within each replication the following regression is estimated:

$$y_{ig} = c + \beta_T Treat_{ig} + \beta_x x_{ig} + \mu_{ig}$$

Where $Treat_{ig}$ is a binary indicator variable of treatment status. The null hypothesis $\beta_T = 0$ is tested using the CRVE procedure assuming the t-statistic follows
a $t(G - 1)$ and $t(G^* - 1)$ distribution. I test this procedure for ten or nine clusters matching the LAD early decision and SLID early decision analysis, and four clusters to match the late decision analysis. The nine and ten cluster designs each have four treated clusters whereas the four cluster design has two treated clusters. In all cases I report the 1%, 5%, and 10% rejection frequencies for each procedure. These results are reported in Table 2.

Results comparing the two distributions when using the CRVE are in the first panel of Table 2. In all cases, as before the rejection frequency when assuming the $t(G - 1)$ distribution is far too large. On the other hand, when assuming the $t(G^* - 1)$ procedure the rejection frequencies are much lower. For the ten cluster design there is slight overrejection at all three levels. For the nine cluster design there is slight underrejection at the 1% and 5% level and overrejection at the 10% level, however these frequencies are quite close to the size of the test. Both these designs however have much more reasonable results than the standard CRVE procedure assuming the $t(G - 1)$ distribution. In the four cluster design, there is still overrejection at the 5% and 10% level when assuming the $t(G^* - 1)$ distribution however the overrejection is much less severe than the standard distribution.

The results from the 6 point bootstrap are in the second panel of Table 2. In all cases, it underrejects severely at the 1% level. It performs worse than the $t(G^* - 1)$ CRVE in the nine cluster design, with overrejection at the 5% and 10% level. In the four cluster design however it works much better at the 5% level. Like the CRVE, the 6 point bootstrap still overrejects severely at the 10% level. In the ten cluster design, it works very slightly better than the CRVE at the 5% and 10% level.

The results from this Monte Carlo suggest that careful attention must be paid to inference. In the early decision analysis, when using data from all ten provinces, the CRVE assuming a $t(G^* - 1)$ distribution will generally result in a well-sized test. However, in the late decision analysis, when using data from only the four Atlantic provinces inference will be more difficult. In this case, the cluster robust $P$-values will be quite significantly understated, while the wild cluster bootstrap $P$-values will be correctly sized at the 5% level but overrejecting at the 10% level.

The CRVE, assuming the $t(G^* - 1)$ distribution, gives the most reasonable results. In the nine cluster case this test is remarkably well-sized. The ten cluster design results in the $P$-values being slightly understated along with the $P$-values being understated at the 5% and 10% level for the four cluster design. For the four cluster design, the 6 point bootstrap gives a well-sized test at the 5% level, while the CRVE
gives a well-sized test at the 1% level. Thus, henceforth unless otherwise stated I report the CRVE P-values assuming the t-statistics follow the $t(G^* - 1)$ distribution for the early decision analysis, and include the 6 point bootstrap results for the late decision analysis. $G^*$ here is estimated using the same Monte Carlo setup where in each replication it is estimated using the clusteff Stata program described in Lee and Steigerwald (2018). I use the average of all the replications which for the ten, nine, and four cluster regressions are estimated to be 5.17, 3.7, and 2.8 respectively.

5 Data

The primary dataset for this paper comes from Statistics Canada’s Longitudinal Administrative Databank (LAD). The dataset uses administrative tax-file data on a 20% sample of Canadians. As the name suggests, the databank is longitudinal and contains tax information for included individuals from 1982–2011. The dataset also records age, gender, and province of residence. The LAD allows for estimation of all of the outcomes of interest. The longitudinal nature of the data allows for a nearly perfect measure of migration. Certain items in the tax code allow a researcher to infer university enrollment. The procedure for this estimation was developed by Morissette, Chan and Lu (2015) and is explained in the Appendix. These same items can be used to infer whether someone graduated from university. Thus, the LAD can be used to estimate mobility, university enrollment, and educational attainment.

The sample used for the first part of the analysis is individuals aged 18 – 23 in the years 2000 – 2011, and the outcomes are summarized in Table 3. Enrolled in University is a binary indicator for whether an individual was enrolled in university during the current tax year. Ever Enrolled in University is a binary indicator for whether that individual was ever enrolled in university. Enrolled in University for 2 years is a binary indicator for whether the individual had been enrolled in university for at least two years, as measured by having been enrolled for 16 full-time equivalent months. University Graduate is a binary indicator for whether the individual graduated from university, measured by the individual having been enrolled for 24 full-time equivalent months. University Dropout is a binary indicator for whether the individual had dropped out of university. The variable is set equal to one for

30 For more information see http://www23.statcan.gc.ca/imdb/p2SV.pl?Function=getSurvey&SDDS=4107.
31 Individuals in Canada are required to file both federal and provincial taxes and are thus required to file with a province of residence in each year.
32 See the appendix for details.
individuals who have previously enrolled in university, have not graduated from university, and are not currently enrolled in university. Student Loan Interest (to date) measures the cumulative total of student loan interest that an individual has claimed on her tax returns. This variable is analyzed only for those who had enrolled in university. Finally, Student Loan Interest (to date) ≥ 0 is the same variable, conditional on claiming a positive amount of interest paid.

Late period decisions are analyzed using data from individuals aged 23 – 28 over the years 2000 – 2011. These variables are summarized in Table 5. Three primary outcomes are analyzed: the university graduation decisions, whether to move out of provinces, and whether to move out of Atlantic Canada. University Graduate is defined in the same way as before. It is a binary outcome defined for those who had ever enrolled in university. It is set equal to one for those who had been enrolled in university for at least 24 months, and set to zero otherwise. Moved Province is a binary indicator for whether an individual had moved out of province, which is analyzed only for university enrollees. The variable is set to one in the year an individual moved out of province, and any subsequent year. The variable is set to zero otherwise. Moved From Atlantic is similarly defined. It is set to one in the year in which an individual moves from the Atlantic Region, and any subsequent year. It is set to zero otherwise.

A secondary dataset used for analysis is the confidential version of the Survey of Labour and Income Dynamics (SLID), which is conducted by Statistics Canada. In contrast to the LAD, the SLID directly asks respondents about education outcomes. Thus, if consistent estimates are found in both samples, we can have faith in the imputation process used in the LAD analysis. The SLID surveys roughly 60,000 individuals each year, with individuals surveyed for six consecutive years. Data is recorded at the individual level. The sample used is observations from the years 2000 – 2010. The sample for the first part of the analysis is individuals aged 18 – 23 from all provinces, except for Quebec. The sample for the second part of the analysis is individuals aged 23 – 28, from the Atlantic provinces.

The analyzed early period variables from the SLID are summarized in Table 7. In general, these variables are constructed to be similar to the variables analyzed in the LAD. University Ever is a binary indicator for whether an individual has ever enrolled in university. University Dropout is a binary indicator for whether an

33 Unfortunately, with the SLID being a survey, there is some attrition. Survey weights are used to correct for attrition.
individual has ever dropped out of university. This variable is set equal to one if the individual had ever enrolled in university, is not currently enrolled in university, and is not a university graduate. College Ever and College Dropout are similarly defined for individuals in college. Finally, Post Sec. Ed. Ever and Post Sec. Ed. Dropout are defined in the same way if the individual had enrolled (dropped out of) college or university.

Table 9 summarizes the late period variables from the SLID. These variables analyze decisions that would have been made by individuals at a slightly later stage of life. Same Prov. as High School is a binary indicator of whether the individual currently lives in the same province as where they attended high school. This variable is set equal to one for individuals in the same province, and set equal to zero otherwise. The variable is also summarized for individuals who had attended university, shown in the second row of the table. University Graduate is a binary indicator for whether an individual had graduated from university. Similarly, College Graduate is a binary indicator for whether an individual had graduated from college, and Post Sec. Ed. Graduate an indicator for college or university graduation.

6 Results

LAD – Early Period Decisions

Results from the early period LAD regressions can be found in Table 4. The results from the regression suggest that the programs had some impacts. In general, the likelihood of individuals enrolling in university was unchanged. The coefficient for Ever Enrolled in University is both very small in magnitude and highly insignificant. Similarly, the coefficient on Enrolled in University for 2 years is not significant. University Dropout is estimated to decrease by a relatively large 4.1%, and the coefficient has a cluster robust $P$-value of 0.050. It is not too surprising that the coefficient on University Graduate is not significant, as the sample here is individuals 18–23. Finally, Student Loans Interest suggests that student loan interest has decreased, however Student Loans Interest > 0 is less statistically significant. These coefficients are both economically significant. The sample for these variables is individuals who had enrolled in university, aged 18–23, many of whom will still be in school.
LAD – Late Period Decisions

Table 6 shows the estimates of the late period decisions using the LAD. The results show that the likelihood of graduating increased for males at the 10% level. The coefficient for University Graduate is estimated to increase by 0.062% for males with some statistical significance. The estimates on migration decisions suggest that migration decisions were not affected. In the combined sample, the estimates for both Moved Province and Moved From Province are statistically insignificant. The estimate for Moved Province for males has a cluster robust $P$-value of 0.076, with a coefficient of 0.029. This is the opposite of the goal of the programs as it suggests individuals were more likely to move. The other coefficients for the migration variables are statistically insignificant. Taken together since no variable is statistically significant at the 5% level, these estimates suggest that the programs were ineffective in retaining university educated individuals within province. This is consistent with the evidence given by the Nova Scotia government when it cancelled its retention program.

SLID – Early Period Decisions

The results of the early period analysis using the SLID data can be found in Table 8. The results for the combined sample are displayed in the top of the table, results for males in the middle, and results for females in the bottom. In general, the results for males and females and the combined sample are quite similar, with the majority of estimates being quite statistically insignificant. One notable exception is the coefficient for University Dropout, which is estimated to decrease by \(-0.049\) in the combined sample. This estimate is statistically significant at 5% using cluster robust standard errors, and significant at the 10% level using the 6-point bootstrap $P$-value. Recall that the $t(G^* - 1)$ CRVE procedure used here is quite well-sized in this nine cluster case. Interestingly, when the impacts are estimated separately by gender, the estimate for males is both larger and of greater statistical significance than for females. In fact, using the CRVE $P$-values the coefficients on University Dropout for the combined sample and for males are the only coefficients significant at the 10% level with bootstrap $P$-values. This suggests that the programs did not have any impact on university or college enrollment, as none of these estimates is remotely statistically significant. In this case we can be confident of the $P$-values as the Monte Carlo results would suggest we have a well-sized test.
SLID – Late Period Decisions

The results of the late period analysis using the SLID can be found in Table 10. These estimates are meant to measure the extent to which the programs influenced graduation rates and migration decisions. The most notable result from the table is that the estimated coefficients on the Same Province variables are negative in all three panels. This is especially interesting as the stated intention of the programs was to keep recent graduates in province. The majority of the coefficients are not remotely statistically significant when using the CRVE or 6-point $P$-values for inference. The one exception is the coefficient on Same Province for males, who had attended university, which is significant at the 10% and 5% level for each procedure. This could suggest that these individuals were less likely to stay in province when the graduate retention programs were on offer. The lack of statistical significance on the other coefficients suggests that these individuals were not any more likely to graduate from either college or university. Again, the 6-point $P$-values can be thought of as a well-sized test at the 5% level for the $P$-value.

6.1 Robustness

As a robustness check I have collected the historical retention rate from the Maclean’s Magazine University Rankings. The university ranking data contains self reported statistics from nearly 50 universities in Canada. The annual nature of the rankings allows for the construction of a panel, for this paper I have a panel covering the years 2002 through 2013.

The Maclean’s variable measures the retention rate for each university in the sample. The retention rate reports the proportion of first year full time students who re-enroll at the same university in the following academic year. This variable is a good proxy for overall graduation rates as the majority of individuals who do not finish university leave before their second year of university. The estimates in Table 11 are presented as both pooled estimates and estimates on a province by province basis.

The estimated impact of the graduate retention programs shows that the retention rate has increased by a reasonably large 2.315 percentage points. However, the statistical significance of this estimate is somewhat wanting, as the 6-point bootstrap

\footnote{I’m very appreciative of Mary Dwyer from Maclean’s for providing me with past editions of the Maclean’s rankings.}

\footnote{See the Globe and Mail focus on ‘Our Time to Lead: Education’ from October 6, 2012.}
$P$-value is 14.3%. The province-specific coefficients in the bottom of the table are interesting, as they suggest that the programs in Manitoba and Saskatchewan had much larger impacts than the programs in Nova Scotia and New Brunswick. Neither the cluster-robust $P$-values nor the 6-point $P$-values are reliable for these coefficients as only one group is ‘treated’.\footnote{See MacKinnon and Webb (2017, 2020, 2018) for more details.}

7 Conclusions

Several provinces in Canada have introduced generous tax credits to increase the local stock of college educated individuals. Although the programs were designed to keep individuals in a province after graduation they explicitly encourage graduation. The impact of these programs on a variety of education and migration decisions is evaluated using both administrative and survey data. The programs have no discernible impact on university enrollment. University dropout rates decrease by 4.1% for individuals aged 18-23. Finally, the programs had no impact on out of province migration for recent graduates.

From a policy perspective, the implications of this analysis are somewhat ambiguous. The credits were introduced to discourage interprovincial migration. While it appears the credits did not change the pattern of migration, they did decrease university dropout rates. If those individuals who did not dropout go on to graduate, then the programs may have met the goal of increasing the average education level in the province. It is possible that the programs are superior to the Merit Scholarships in that they both offer financial relief for local students, and directly encourage individuals to graduate. Past research has also found the Merit programs to be distortionary. Cohodes and Goodman (2014) find college completion rates to decrease. Sjoquist and Winters (2013a) find that merit scholarships decrease the likelihood that a student majored in a STEM field. Future research should examine whether the retention credits had similar distortionary effects.

The credits are also perhaps a better alternative to in-state tuition, as they effectively reduce the net tuition for those who stay in province. Aghion et al. (2009) argue that investments in educational institutions in states far away from the technology frontier tend to benefit states closer to the frontier, as graduates of said institutions tend to migrate toward the frontier. Thus, it is perhaps preferable to offer reduced tuition for those who stay in province, as opposed to those who attend
school in province. Whether this change would be regressive in its incidence is beyond the scope of this paper, though it is encouraging that the credits reduced the amount paid in student loan interest. While the programs appear to encourage those enrolled in university to finish their education, the costly nature of these programs makes further experimentation difficult.
References


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Neill, Christine (2013) ‘What you don’t know can’t help you: Lessons of behavioural economics for tax-based student aid.’ *CD Howe Institute Commentary*


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Sjoquist, David L., and John V. Winters (2013a) ‘State Merit-Aid Programs and College Major: A Focus on STEM.’ IZA Discussion Papers 7381, Institute for the Study of Labor (IZA), May


Webb, Matthew D. (2014) ‘Reworking wild bootstrap based inference for clustered errors.’ Working Papers 1315, Queen’s University, Department of Economics


Table 1: Summary of the various Graduate Retention Programs

<table>
<thead>
<tr>
<th></th>
<th>SK</th>
<th>MB</th>
<th>NS</th>
<th>NB</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>maximum amount</strong></td>
<td>20k</td>
<td>25k</td>
<td>15k</td>
<td>20k</td>
</tr>
<tr>
<td><strong>rebate per year</strong></td>
<td>10%,20%</td>
<td>4k, 10%</td>
<td>2.5k</td>
<td>4k</td>
</tr>
<tr>
<td><strong>NPV ($000) @ 5%</strong></td>
<td>16.9</td>
<td>14.1</td>
<td>13.3</td>
<td>12.6</td>
</tr>
<tr>
<td><strong>refundable credit</strong></td>
<td>Y*</td>
<td>N</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td><strong>roll-over credit</strong></td>
<td>N*</td>
<td>Y</td>
<td>N</td>
<td>Y</td>
</tr>
<tr>
<td><strong>eligibility duration</strong></td>
<td>7</td>
<td>10</td>
<td>6</td>
<td>20</td>
</tr>
<tr>
<td><strong>application req.</strong></td>
<td>N</td>
<td>N</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td><strong>tuition based</strong></td>
<td>Y</td>
<td>Y</td>
<td>N</td>
<td>Y</td>
</tr>
<tr>
<td><strong>tuit. % refunded</strong></td>
<td>100%</td>
<td>60%</td>
<td>-</td>
<td>50%</td>
</tr>
<tr>
<td><strong>program costs</strong></td>
<td>35m</td>
<td>34m</td>
<td>25m</td>
<td></td>
</tr>
</tbody>
</table>

Notes: * reflects the change in 2012 that Saskatchewan announced, where the credit was no longer refundable, but would instead roll-over from one year to the next. The NPV calculation assumes a 5% discount rate, sufficient earnings to get the maximum credit in all years, and $22,663 in tuition paid in earning a 4 year B.A. degree from Queen’s University.

Table 2: Treatment Model Monte Carlo Results

<table>
<thead>
<tr>
<th></th>
<th>10 Clusters</th>
<th>9 Clusters</th>
<th>4 Clusters</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>CRVE</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>t(G-1)</td>
<td>3.804%</td>
<td>6.024%</td>
<td>8.224%</td>
</tr>
<tr>
<td>1% rej. freq.</td>
<td>1% rej. freq.</td>
<td>1.200%</td>
<td>0.568%</td>
</tr>
<tr>
<td>5% rej. freq.</td>
<td>11.464%</td>
<td>15.468%</td>
<td>22.924%</td>
</tr>
<tr>
<td>10% rej. freq.</td>
<td>19.096%</td>
<td>23.724%</td>
<td>33.612%</td>
</tr>
<tr>
<td><strong>6 pt. Boot</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1% rej. freq.</td>
<td>0.180%</td>
<td>0.644%</td>
<td>0.176%</td>
</tr>
<tr>
<td>5% rej. freq.</td>
<td>5.504%</td>
<td>7.924%</td>
<td>4.900%</td>
</tr>
<tr>
<td>10% rej. freq.</td>
<td>11.868%</td>
<td>14.836%</td>
<td>17.732%</td>
</tr>
</tbody>
</table>

Notes: Results from 25,000 Monte Carlo simulations. Average G* for the 10, 9, and 4 cluster designs across all replications were 5.17, 3.7, and 2.8 respectively.
Table 3: LAD - Summary statistics - Early Period

<table>
<thead>
<tr>
<th></th>
<th>GRP Pre</th>
<th>GRP Post</th>
<th>Not GRP Pre</th>
<th>Not GRP Post</th>
</tr>
</thead>
<tbody>
<tr>
<td>Enrolled in University</td>
<td>.149</td>
<td>.229</td>
<td>.108</td>
<td>.146</td>
</tr>
<tr>
<td>Ever Enrolled in University</td>
<td>.448</td>
<td>.372</td>
<td>.39</td>
<td>.278</td>
</tr>
<tr>
<td>Enrolled in University for 2 years</td>
<td>.751</td>
<td>.607</td>
<td>.788</td>
<td>.583</td>
</tr>
<tr>
<td>University Graduate</td>
<td>.622</td>
<td>.417</td>
<td>.639</td>
<td>.348</td>
</tr>
<tr>
<td>University Dropout</td>
<td>.309</td>
<td>.27</td>
<td>.29</td>
<td>.316</td>
</tr>
<tr>
<td>Student Loan Interest (to date)</td>
<td>220</td>
<td>20</td>
<td>110</td>
<td>10</td>
</tr>
<tr>
<td>Student Loan Interest (to date) &gt; 0</td>
<td>300</td>
<td>500</td>
<td>920</td>
<td>270</td>
</tr>
</tbody>
</table>

Table 4: LAD - Regression Results - Early Period

<table>
<thead>
<tr>
<th></th>
<th>Coeff.</th>
<th>t(G-1)</th>
<th>t(G*-1)</th>
<th>Obs</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ever Enrolled In University</td>
<td>-0.00098</td>
<td>0.975</td>
<td>0.976</td>
<td>458692x</td>
</tr>
<tr>
<td>Enrolled in University for 2 years</td>
<td>0.015444</td>
<td>0.452</td>
<td>0.476</td>
<td>171163x</td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.017156</td>
<td>0.156</td>
<td>0.197</td>
<td>171163x</td>
</tr>
<tr>
<td>University Dropout</td>
<td>-0.04111</td>
<td>0.021</td>
<td>0.050</td>
<td>171163x</td>
</tr>
<tr>
<td>Student Loan Interest (to date)</td>
<td>-15.7603</td>
<td>0.001</td>
<td>0.009</td>
<td>171163x</td>
</tr>
<tr>
<td>Student Loan Interest (to date) &gt; 0</td>
<td>-42.8224</td>
<td>0.029</td>
<td>0.061</td>
<td>10208x</td>
</tr>
</tbody>
</table>

Note: Reported are the P-values from the CRVE assuming the two above distributions for the t-statistics. Sample consists of individuals aged 18-23 from 2000-2011 in all Provinces. Treatment Defined by Year when 18. 5.17 is used as the G* here.

Table 5: LAD - Summary Statistics - Late Period

<table>
<thead>
<tr>
<th></th>
<th>GRP Pre</th>
<th>GRP Post</th>
<th>Not GRP Pre</th>
<th>Not GRP Post</th>
</tr>
</thead>
<tbody>
<tr>
<td>Female</td>
<td>0.512</td>
<td>0.5</td>
<td>0.496</td>
<td>0.5</td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.587</td>
<td>0.652</td>
<td>0.52</td>
<td>0.538</td>
</tr>
<tr>
<td>Moved</td>
<td>0.458</td>
<td>0.322</td>
<td>0.529</td>
<td>0.32</td>
</tr>
<tr>
<td>Moved From Atlantic</td>
<td>0.423</td>
<td>0.301</td>
<td>0.497</td>
<td>0.302</td>
</tr>
<tr>
<td>Obs</td>
<td>197000</td>
<td>86800</td>
<td>101400</td>
<td>22000</td>
</tr>
</tbody>
</table>
Table 6: LAD - Regression Results - Late Period

<table>
<thead>
<tr>
<th>Males and Females</th>
<th>Coef</th>
<th>t(G-1)</th>
<th>t(G*-1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>University Graduate</td>
<td>0.034</td>
<td>0.039</td>
<td>0.107</td>
</tr>
<tr>
<td>Moved Province</td>
<td>0.009</td>
<td>0.662</td>
<td>0.688</td>
</tr>
<tr>
<td>Moved From Atlantic</td>
<td>0.001</td>
<td>0.917</td>
<td>0.923</td>
</tr>
<tr>
<td><strong>Males</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.062</td>
<td>0.030</td>
<td>0.093</td>
</tr>
<tr>
<td>Moved Province</td>
<td>0.029</td>
<td>0.020</td>
<td>0.076</td>
</tr>
<tr>
<td>Moved From Atlantic</td>
<td>0.013</td>
<td>0.062</td>
<td>0.137</td>
</tr>
<tr>
<td><strong>Females</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.014</td>
<td>0.380</td>
<td>0.438</td>
</tr>
<tr>
<td>Moved Province</td>
<td>-0.011</td>
<td>0.707</td>
<td>0.729</td>
</tr>
<tr>
<td>Moved From Atlantic</td>
<td>-0.008</td>
<td>0.144</td>
<td>0.225</td>
</tr>
</tbody>
</table>

**Note:** Reported are the $P$-values from the CRVE assuming the two above distributions for the t-statistics. Sample consists of individuals aged 23-28 from 2000-2011 in all Atlantic Provinces. Treatment Defined by Year when 23. 2.8 is used as the $G^*$ here.

Table 7: SLID - Summary statistics - All Provinces

<table>
<thead>
<tr>
<th></th>
<th>Pre</th>
<th>Post</th>
<th>Not Pre</th>
<th>Not Post</th>
</tr>
</thead>
<tbody>
<tr>
<td>University Ever</td>
<td>0.442</td>
<td>0.404</td>
<td>0.397</td>
<td>0.384</td>
</tr>
<tr>
<td>University Dropout</td>
<td>0.089</td>
<td>0.022</td>
<td>0.090</td>
<td>0.037</td>
</tr>
<tr>
<td>College Ever</td>
<td>0.688</td>
<td>0.634</td>
<td>0.649</td>
<td>0.635</td>
</tr>
<tr>
<td>College Dropout</td>
<td>0.111</td>
<td>0.061</td>
<td>0.119</td>
<td>0.051</td>
</tr>
<tr>
<td>Post Sec. Ed. Ever</td>
<td>0.319</td>
<td>0.280</td>
<td>0.349</td>
<td>0.357</td>
</tr>
<tr>
<td>Post Sec. Ed. Dropout</td>
<td>0.129</td>
<td>0.113</td>
<td>0.135</td>
<td>0.072</td>
</tr>
</tbody>
</table>
Table 8: SLID - School Enrollment and Dropout Decisions

<table>
<thead>
<tr>
<th></th>
<th>Males and Females</th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Coeff.</td>
<td>t(G-1)</td>
<td>t(G*-1)</td>
<td>6-point</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>OLS</td>
<td>CRVE</td>
<td>CRVE</td>
<td>Boot</td>
<td>Obs</td>
</tr>
<tr>
<td>University Ever</td>
<td>-0.007</td>
<td>0.741</td>
<td>0.840</td>
<td>0.850</td>
<td>0.884</td>
<td>37848</td>
</tr>
<tr>
<td>University Dropout</td>
<td>-0.049</td>
<td>0.004</td>
<td>0.002</td>
<td>0.029</td>
<td>0.087</td>
<td>13226</td>
</tr>
<tr>
<td>College Ever</td>
<td>0.006</td>
<td>0.760</td>
<td>0.781</td>
<td>0.795</td>
<td>0.890</td>
<td>37964</td>
</tr>
<tr>
<td>College Dropout</td>
<td>-0.003</td>
<td>0.822</td>
<td>0.517</td>
<td>0.553</td>
<td>0.660</td>
<td>24396</td>
</tr>
<tr>
<td>Post Sec. Ed. Ever</td>
<td>0.016</td>
<td>0.419</td>
<td>0.422</td>
<td>0.468</td>
<td>0.613</td>
<td>37588</td>
</tr>
<tr>
<td>Post Sec. Ed. Dropout</td>
<td>0.046</td>
<td>0.054</td>
<td>0.002</td>
<td>0.029</td>
<td>0.111</td>
<td>13728</td>
</tr>
</tbody>
</table>

Males

<table>
<thead>
<tr>
<th></th>
<th>Coeff.</th>
<th>t(G-1)</th>
<th>t(G*-1)</th>
<th>6-point</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>University Ever</td>
<td>0.036</td>
<td>0.226</td>
<td>0.255</td>
<td>0.320</td>
<td>0.408</td>
<td>18592</td>
</tr>
<tr>
<td>University Dropout</td>
<td>-0.075</td>
<td>0.007</td>
<td>0.004</td>
<td>0.038</td>
<td>0.092</td>
<td>5269</td>
</tr>
<tr>
<td>College Ever</td>
<td>0.032</td>
<td>0.269</td>
<td>0.139</td>
<td>0.213</td>
<td>0.351</td>
<td>18654</td>
</tr>
<tr>
<td>College Dropout</td>
<td>-0.015</td>
<td>0.520</td>
<td>0.538</td>
<td>0.572</td>
<td>0.712</td>
<td>10877</td>
</tr>
<tr>
<td>Post Sec. Ed. Ever</td>
<td>0.000</td>
<td>0.994</td>
<td>0.994</td>
<td>0.994</td>
<td>0.993</td>
<td>18452</td>
</tr>
<tr>
<td>Post Sec. Ed. Dropout</td>
<td>0.030</td>
<td>0.389</td>
<td>0.352</td>
<td>0.406</td>
<td>0.548</td>
<td>6671</td>
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</tbody>
</table>

Females

<table>
<thead>
<tr>
<th></th>
<th>Coeff.</th>
<th>t(G-1)</th>
<th>t(G*-1)</th>
<th>6-point</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>University Ever</td>
<td>-0.053</td>
<td>0.073</td>
<td>0.284</td>
<td>0.346</td>
<td>0.395</td>
<td>19256</td>
</tr>
<tr>
<td>University Dropout</td>
<td>-0.031</td>
<td>0.145</td>
<td>0.316</td>
<td>0.374</td>
<td>0.431</td>
<td>7957</td>
</tr>
<tr>
<td>College Ever</td>
<td>-0.028</td>
<td>0.277</td>
<td>0.385</td>
<td>0.435</td>
<td>0.545</td>
<td>19310</td>
</tr>
<tr>
<td>College Dropout</td>
<td>0.003</td>
<td>0.884</td>
<td>0.882</td>
<td>0.889</td>
<td>0.911</td>
<td>13519</td>
</tr>
<tr>
<td>Post Sec. Ed. Ever</td>
<td>0.023</td>
<td>0.394</td>
<td>0.436</td>
<td>0.481</td>
<td>0.621</td>
<td>19136</td>
</tr>
<tr>
<td>Post Sec. Ed. Dropout</td>
<td>0.049</td>
<td>0.111</td>
<td>0.214</td>
<td>0.283</td>
<td>0.332</td>
<td>7057</td>
</tr>
</tbody>
</table>

**Note:** Reported are the P-values for the treatment for any given outcome variable. Sample consists of individuals aged 18-23 from 2000-2010 in Provinces except Quebec. Treatment Defined by Year when 18. 3.7 is used as the G* here.

Table 9: SLID - Summary statistics - Atlantic Provinces

<table>
<thead>
<tr>
<th></th>
<th>GRP</th>
<th>Not GRP</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Pre</td>
<td>Post</td>
</tr>
<tr>
<td>Same Prov as High School</td>
<td>0.776</td>
<td>0.766</td>
</tr>
<tr>
<td>&quot;</td>
<td>Attended University</td>
<td>0.765</td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.354</td>
<td>0.275</td>
</tr>
<tr>
<td>College Graduate</td>
<td>0.583</td>
<td>0.542</td>
</tr>
<tr>
<td>Post Sec. Ed. Graduate</td>
<td>0.543</td>
<td>0.474</td>
</tr>
</tbody>
</table>

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### Table 10: SLID - Migration and Graduation Decisions

<table>
<thead>
<tr>
<th></th>
<th>Males and Females</th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Coeff. OLS CRVE CRVE Boot Obs</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Same Prov as High School</strong></td>
<td>-0.062 0.190 0.343 0.405 0.371 9324</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&quot; &quot;</td>
<td>Attended University</td>
<td>-0.113 0.031 0.053 0.126 0.172 7324</td>
<td></td>
<td></td>
</tr>
<tr>
<td>University Graduate</td>
<td>0.049 0.382 0.457 0.506 0.757 3959</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>College Graduate</td>
<td>0.059 0.307 0.515 0.557 0.652 4958</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Post Sec. Ed. Graduate</td>
<td>0.061 0.168 0.314 0.380 0.558 7324</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Note:** Reported are the P-values for the treatment for any given outcome variable.
Sample consists of individuals aged 23-28 from 2000-2010 in all Atlantic Provinces.
Treatment Defined by Year when 23. 3.7 is used as the G* here.

### Table 11: Maclean’s - Percentage of Students Returning for 2nd Year

<table>
<thead>
<tr>
<th></th>
<th>Coeff. OLS CRVE CRVE Boot</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>All</strong></td>
<td>2.315 0.021 0.053 0.143</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SK</td>
<td>4.594 0.000 0.000 0.622</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MB</td>
<td>4.006 0.000 0.000 0.530</td>
<td></td>
<td></td>
</tr>
<tr>
<td>NS</td>
<td>2.141 0.000 0.001 0.320</td>
<td></td>
<td></td>
</tr>
<tr>
<td>NB</td>
<td>-1.009 0.043 0.074 0.550</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Notes:** Estimates Clustered by Province
The procedure used by Morissette, Chan and Lu (2015) (MCL, henceforth) to estimate university enrollment is described in detail in the appendix of their paper. This procedure borrows heavily from theirs. Since 1999 the tax code has the following three variables which allow for inference of university enrollment rates:

- tuition fees for self
- educational deduction for full-time students
- educational deduction for part-time students

MCL use these variables to construct enrollment rates, the procedure is as follows:

1. I take the total education deduction for full time students and divide by the maximum amount allowed per month by the Canada Revenue Agency. This gives the total number of months the individual was enrolled full-time in college or university that year.

2. A similar calculation is performed for the part-time months.

3. An estimate of the months of full-time-equivalent enrollment is derived. Full-time months are counted as 1.0 months and part-time months are counted as 0.6 months.

4. The amount claimed in tuition is divided by the number of months estimated in step 3. This yields an estimated tuition per month.

5. The monthly average tuition amount is multiplied by 8, to compute a full-time-equivalent annual amount.

6. The annual amount from step 5 is compared to 0.8 – 2.0 times the annual provincial average undergraduate tuition amount. If the annual amount is

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37 I am grateful to René Morissette for providing me with SAS code to implement the procedure.
38 The maximum amounts come from either Morissette, Chan and Lu (2015) or Neill (2013).
39 To correctly estimate the average tuition, the amount reported is converted into a full-time-equivalent tuition amount.
40 The values used are two year averages, owing to the overlap of the school year and the tax year. The tuition values used from 1977–2011 come from the following sources: the values from 1977–2000 come from Johnson and Rahman (2005), I thank David R. Johnson for providing me with these values; 2001–2008 come from MCL; 2009–2011 from Statistics Canada Table B.2.9 http://www.statcan.gc.ca/pub/81-562-x/2014001/tbl/12b2.9-eng.htm.
within the interval, then the variable “university enrolled” is set equal to 1, it is set equal to 0 otherwise.\footnote{Unfortunately, for the purposes of this study, the tax system does not differentiate between college and university enrollment. The MCL procedure uses the fact that university tuition is much more costly than college tuition to infer university attendance. In 2013 average undergraduate university tuition in Canada was $5,772 (see statscan table referenced above), while average college tuition in Ontario was $1,900 (http://www.tcu.gov.on.ca/pepg/audiences/colleges/costs_coll.html#tuition).}

Building on the MCL procedure, I take advantage of the panel nature of the dataset to construct additional variables. I sum the months variable from step three over all the years in the dataset, to estimate the total months spent in university. I then compare this variable to a number of key thresholds. For “persistence”, I estimate whether the individual completed two or more years of university, by calculating whether the total number of months exceeded 16.

Similarly, I estimate whether someone graduated from university by comparing the total months variable to 24. University programs are typically 32 months in length in most of Canada, but programs in Quebec and non-honours undergraduate programs can be 24 months in length.\footnote{The majority of dropouts occur early in university careers. Stinebrickner and Stinebrickner (2013) report that of those who dropped out –of one particular college – 40% dropped out in year 1, 34.4% in year 2, and 25.1% in year 3. Similarly, the Maclean’s graduation rate is on average 86.5% of the retention rate, suggesting that the largest fraction of dropouts leave after the first year. Thus the majority of those who finished 24 months of university will be university graduates.}

Finally, university dropout is set equal to 1 if an individual was a) not already a university graduate, b) had previously been enrolled in university, and c) was not currently enrolled in university.
Declaration of Interest

I have received funding in excess of $10,000 from both the Social Sciences and Humanities Research Council (Canada) and from my employer. Neither source of funding has any oversight of my research output. I have no other conflicts of interest to declare.