

How Low-Income Expectations Affect Student Loan Repayment Plan Choice: Survey Evidence from College Seniors¹

By Joshua Brownstein

Updated: January 2023

Abstract

Income-driven repayment plans lower required payments for student loan borrowers when their income decreases. This helps to reduce student loan defaults. Despite universal availability, only a minority of student loan borrowers in the U.S. are in an income-driven repayment plan. In this study, I test whether a student's choice of repayment plan is related to their expectations of earning a low income. Using an information experiment in a web survey, I create two groups of college seniors with an exogenous difference in low-income expectations. I find that respondents who see the major specific income information believe they, on average, have a higher probability of earning a low income. However, those respondents are not any more likely to choose the income-driven repayment plan. I conclude that students' repayment plan preferences are not strongly related to their expectations of earning a low income. This may be due to students caring about things other than minimizing monthly payments when choosing a repayment plan.

¹ Thank you to Scott Imberman, Ajin Lee, Leslie Papke, and Kris Renn for providing me feedback on this project as members of my dissertation committee. Thank you to Cody Orr for providing me code in Qualtrics for creating the income expectations questions. Thank you to the National Science Foundation (NSF) and the Michigan State University College of Social Science for funding this research. Thank you to Patria Wilson for helping me to fill out the NSF grant for this project. Thank you to Belen Freight for helping me use the NSF grant to incentivize students to take the survey. Thank you to the economics doctoral students at Michigan State University who provided me feedback on this project during graduate student seminars.

I. Introduction and Motivation

Most college students in the United States (U.S.) get loans from the federal government to fund their college education (Woo, Bentz, Lew, Velez, and Smith 2017). The U.S. federal government offers student loan borrowers a choice between two kinds of repayment plans. One type of repayment plan sets payments so that the loan is paid off within a certain period². The other type of repayment plan sets payments as a function of a borrower's income³. The latter kind of plan is referred to as an income-driven repayment plan or IDR plan. IDR plans are preferred over time-based plans by scholars of student loans for their ability to reduce the loan payments of student loan borrowers when their incomes are low (Chapman and Dearden 2017).

Borrowers on IDR plans are more likely to make required on time payments (Herbst 2023) and less likely to default on their student loans⁴ (U.S. Government Accountability Office 2015; Muller and Yannelis 2019). Preventing student loan default is important because defaulting on U.S. government student loans can lead to a variety of negative consequences for the borrower. These consequences include: a reporting of the default to credit bureaus leading to reduced access to private sources of credit, collection fees, wage garnishment, the garnishment of the borrower's tax refund, and the inability to get more U.S. government student loans until the default is resolved. As of Q2 2021, 17 percent of student loan borrowers were in default (Ma and Pender, 2021). Scott-Clayton (2019) finds that the proportion of students who graduated in 1996 who had ever defaulted on their loans continued to increase over the 20 years they had data for. Using that data to forecast

² The default plan choice for students who get loans from the U.S. federal government sets payments so that the loans are paid off in full if the minimum payment is made every month for 10 years.

³ After a certain number of years of making payments on one of these plans, all remaining loan balances are forgiven and required payments decrease to \$0. The income-driven plan that is available to all new borrowers of U.S. government student loans, the Revised Pay as You Earn Plan, offers loan forgiveness after 20 years for undergraduate borrowers and after 25 years for graduate borrowers. See <https://studentaid.gov/manage-loans/repayment/plans/income-driven> for more information.

⁴ The U.S. Department of Education defines student loan default as not making required payments for at least 270 days.

defaults in the future, Scott-Clayton projects that 40% of borrowers who graduated college in 2004 would default on their student loans at some point by 2023. Despite these facts, only 32% of borrowers in FY 2021 were in IDR plans (Ma and Pender, 2021). Given the high default rate on student loans, and the fact that IDR plans likely reduce student loan default, it seems as if borrowers' lives could be significantly improved if more of them were on IDR plans.

One reason why there may be both relatively low enrollment in IDR plans, and a high default rate on student loans, is that students have overly optimistic expectations about their future earnings⁵. In terms of reducing required payments, the biggest benefit of being on an IDR plan occurs when a borrower's income is low. If borrowers believe they have an unreasonably low probability of earning a low income, then they may also believe that it is unlikely they will experience reduced payments should they choose an IDR plan instead of a time-based repayment plan. Should a student loan borrower earn a low income after they graduate while being on a time-based repayment plan, their required payments may be such a large proportion of their income that they are unable or unwilling to make them. If this is the case, then presenting students with relevant information about post-college incomes should cause them to: increase the probability that they believe they will earn a low income, be more likely to choose an IDR plan over a time-based repayment plan and reduce the probability that they default on their loans.

⁵ Colon (2021) finds that, on average, a sample of undergraduates at The Ohio State University underestimate the mean salary for employed workers in Ohio age 30 to 50 with specific groups of majors. In Cox, Kreisman, and Dynarski (2020) college students who participated in a laboratory experiment expect the typical earnings of the typical graduate to be \$34,500 while the average earnings of 24-year-old graduates in 2015 was about \$22,000. In a survey of NYU students Wiswall and Zafar (2015a) find that when they asked what NYU undergraduates thought 30 years old college graduates with broad categories of majors in the US earned, the average response is statistically significantly above the authors calculations of the actual population earnings. The authors also find substantial heterogeneity in errors, with many students underestimating population earnings. Betts (1996) finds that in a sample of undergraduates at UC San Diego, the mean beliefs about the average salary of BA holders in 1990 is close to correct although the mean salary of BA's with psychology degrees is statistically significantly below mean beliefs about the salary of psychology graduates.

The purpose of this research is to learn about the effect that student's expectations of earning a low income have on their choice of student loan repayment plan⁶. To study this, I field an online survey to college seniors at Michigan State University (MSU). Survey respondents are asked about the probability they expect to earn an income in different income ranges. They are also asked if they would prefer an IDR or non-IDR (time-based) plan if they had \$30,000 in student loan debt. The survey includes an information experiment where respondents were randomly shown either information about the average income of U.S. college graduates (All Graduates Income Treatment), or information about the median earnings of MSU graduates with majors like their own major (Major Specific Income Treatment). The goal of providing this information is to create an exogenous difference in low-income expectations between respondents who saw the two types of income information. Questions about income expectations and repayment plan choice are asked both before and after the income information is shown.

Controlling for pre-treatment differences in low-income expectations, I find that survey respondents who see the Major Specific Income Treatment have a subjective probability of earning a low income that is a statistically significant 7 percentage points higher than the survey respondents who see the All-Graduates Income Treatment. However, controlling for pre-treatment differences in repayment plan choice, I find that survey respondents who see the Major Specific Income Treatment are an insignificant 2 percentage points less likely to choose the IDR plan. Based on this, and similar results for various subsamples, I conclude that repayment plan choice is not very responsive to changes in low-income

⁶ Abraham, Filiz-Ozbay, Ozbay, and Turner (2020) and Brownstein (2020) find that the probability a student believes they will earn a low-income 6 months after leaving school is statistically significantly correlated with student loan repayment plan choice.

expectations. This may be because students care about things other than minimizing required monthly payments when picking a repayment plan.

II. Background on Student Loans and Income-Driven Repayment in the United States

About 92% of all student loan debt in the U.S. is owed to the U.S. Federal Government (Peter G. Peterson Foundation 2021)⁷. Students who attend college apply for federal loans by filling out the Free Application for Federal Student Aid. Loans are offered to students as part of their overall financial aid package for a university. Students can borrow up to the lesser of either the cost of attendance, or a limit that is based on year in school and dependency status (Kirkham 2020). For federal student loans, there are limits both on the amount of borrowing per year and the lifetime amount of borrowing⁸. In the academic year 2020-2021, 25 percent of undergraduate students borrowed loans directly from the federal government (Ma and Pender 2021). Ma and Pender also found that 55 percent of students who graduated from public and non-profit 4-year universities in the 2019 – 2020 academic year had student loan debt. They calculate that the average amount of debt among people who graduated with debt that year was \$28,400.

One of the major benefits of IDR plans is that they reduce the probability that borrowers default on their student loans. Borrowers⁹ will be current on their loans if they make at least the minimum monthly loan payment. The minimum monthly loan payment is generally¹⁰ determined by the repayment plan the borrower is on. Once a borrower misses a payment, they are considered delinquent on that loan.

⁷ Every year most new student loan debt is also owed to the U.S. Federal Government. For example, in the 2020 – 2021 academic year, 87% of new student loan debt was owed to the Federal Government (Ma and Pender 2021). The other 13% was owed to private companies.

⁸ If students would like to borrow more than the limits for those loans, their parents may borrow Parent's PLUS loans from the federal government up to the cost of attendance.

⁹ In this paper borrowers is used as a shorthand for U.S. citizens who have gotten student loans from the U.S. federal government.

¹⁰ Borrowers can temporarily lower their minimum monthly payment to \$0 using deferment or forbearance. Deferment and forbearance can be given for a variety of approved circumstances such as getting treated for cancer or serving in the Peace Corps.

Borrowers who are delinquent on their loans for a period of 90 days have their delinquency reported to the 3 major Credit Reporting Agencies (CRA's)¹¹. If a borrower pays less than the minimum payment for 270 days, then their loan is in default. Default has several negative consequences for the borrower including: the entire amount of the loan is due immediately, the default is reported to the 3 major CRA's, being charged for collection costs, being prohibited from receiving additional federal student aid until the default is resolved, and sometimes having their wages, tax refunds and federal benefits garnished. To prevent these harms to borrowers, it is a worthwhile goal to reduce student loan defaults.

In the survey, respondents are given the choice between an IDR plan and a non-IDR plan. These plans are based on two¹² of the repayment plans borrowers can choose from when they enter repayment. The non-IDR plan is based on the Standard Repayment Plan. The Standard Repayment Plan sets minimum monthly payments so that the loan would be paid off if the minimum payment is made every month for 10 years. If a borrower does not select a repayment plan before they begin paying back their loans, they are automatically put on the Standard Repayment Plan. The IDR plan is based on the Revised Pay as You Earn Plan (REPAY). Unless a borrower has an FFEL loan, they can get on REPAY¹³. REPAY sets minimum payments equal to 10% of discretionary income with loan forgiveness¹⁴ after 20 years of payments for an undergraduate borrower or 25 years of payments for individuals who borrowed for graduate or professional school.

¹¹ Credit Reporting Agencies (CRA's) are businesses that collect information about people's use of credit and sell that information to third parties (Irby 2020).

¹² There are currently seven different repayment plans for student loans. Four of those repayment plans set minimum payments as a function of the borrower's income.

¹³ <https://fcaa.org/student-loan-repayment-plans/revised-pay-as-you-earn-repaye/>. FFEL stands for Federal Family Education Loan program. These loans, which were available until 2010, were made by private institutions and guaranteed by the federal government.

¹⁴ According to Student Borrower Protection Center (2021), despite the first IDR plan becoming available in the U.S. in 1995, only 32 U.S. student loan borrowers have ever received loan forgiveness because they had been in an IDR plan for a long period of time. Despite this, it is probably the case that loan forgiveness is a salient feature of IDR plans. In Brownstein (2020), I find that decreasing the number of the years until loan forgives for an IDR plan from 20 years to 15 increases the probability surveyed MSU students prefer an IDR plan to a non-IDR plan by about 20 percentage points.

Discretionary income is defined as income above 150% of the federal poverty line¹⁵.

IDR plans lower the required payment of student loan borrowers when their income is low. This is the feature of IDR plans that probably lower a borrower's probability of default. Even if this feature does not prevent defaults, it prevents students from losing a high proportion of their income on student loan payments when they most need the money. These benefits of IDR plans should make IDR plans more attractive to borrowers who believe they are more likely to earn a low-income. However, IDR plans are not always better than non-IDR plans. If an IDR plan successfully lower a borrower's payments, they cause the borrower to accrue more interest on their loan¹⁶ and take longer to pay off their loan. If borrowers care more about that than the benefits of lower payments, then they may continue to prefer a non-IDR plan even if they believe they are more likely to earn a low-income.

Borrowers can learn about student loan repayment, including payment amounts and what repayment plans are available, by doing student loan exit counseling. Most exit counseling is done through a website created by the U.S. Department of Education (DoE)¹⁷. The information I provide students in the survey is like the information borrowers get on the exit counseling website. DoE requires colleges to have borrowers complete student loan exit counseling when they leave school¹⁸. If colleges do not offer exit counseling, they may lose access to federal financial aid (Klepfer, Ferandez, Fletcher, and Webster 2015). Exit counseling provides

¹⁵ For all states except Hawaii and Alaska, 150% of the federal poverty line for a household with a single individual in 2020 was \$19,140 and for a household with 2 individuals was \$25,860. See Office of the Assistant Secretary for Planning and Evaluation (2019).

¹⁶ A borrower may make lower total interest payments when making lower monthly payments if the borrower makes payments long enough for a significant proportion of their loan balance to be forgiven.

¹⁷ <https://studentaid.gov/app/counselingInstructions.action?counselingType=exit>. Schools can do other things to fulfill the requirement to provide exit counseling. However, anyone can use the U.S. Department of Education's website and most schools (including MSU) refer their students to the website for exit counseling.

¹⁸ It may be the case that a large proportion of borrowers do not complete exit counseling. In a survey of 13,000 high debt borrowers, 40% of respondents reported they did not receive any form of student loan counseling (Whitsett and O'Sullivan 2012).

information on loan balances, repayment obligations, and which repayment plans are available to the borrower. During exit counseling borrowers can enter their estimated future income, future expenses, and amount federal. The website then provides students with an estimated initial monthly payment, an estimated total amount paid, and a repayment period of either the number of years in repayment or the number of years until loan forgiveness. As part of this process, borrowers are asked to select a repayment plan from a menu of available repayment plans. The selected plan is sent to the borrower's loan servicer to determine if they are eligible for the plan. If borrowers do not go through exit counseling, or they do not choose a specific repayment plan at the end of exit counseling, they are put on The Standard Repayment Plan. Students can change their repayment plan at any time by contacting their student loan servicer (Lane, 2020)¹⁹.

III. Literature Review

There are many studies which look at the effect of providing students with information about what they can expect to earn after college on decisions related to college. Wiswall and Zafar (2015a) look at how U.S. students change their income expectations after being informed about the earnings of different groups of individuals. Treatments included being shown information about the average income of all college graduates and the average income of college graduates conditional on gender and major. In a companion paper, Wiswall and Zafar (2015b) use the same data to study how changes to major-specific earnings expectations caused by seeing major specific earnings information changed students' expectations of what they would major in. Baker, Bettinger, Jacob, and Marinescu (2018) study the impact of income information on major choice for community college students. Hastings, Neilson, and Zimmerman (2018) find that Chilean student loan applicants who receive information about college- and major- specific

¹⁹ A loan servicer is a private company that the U.S. federal government contracts with to collect federal student loan payments.

incomes of past Chilean college graduates are less likely to attend and believe they would earn less if they enrolled in, programs whose graduates earned low incomes. Bleemer and Zafar (2018) find that providing information to U.S. household heads about the expected returns to college increase the probability that respondents said they wanted to attend college. Hurwitz and Smith (2018) look at the effect of the release of a large amount of information about the income of college graduates in the College Scorecard. They find that after the information was released colleges with higher reported median incomes had more students send their SAT scores to them. Conlon (2021) finds that students are more likely to choose a major which they received income information about in an online survey. The above research shows that college students change their expectations and behaviors in response to seeing information on post-college incomes.

Another group of studies uses experiments to study what affects student loan repayment plan choice. Abraham, Filiz-Ozbay, Ozbay, and Turner (2020) study how the description of IDR plans affects repayment plan choice. They find that students were statistically significantly more likely to choose the IDR plan when the description of the plan emphasizes its benefits. Cox, Kreisman, and Dynarski (2020) have college students participate in an incentivized laboratory experiment which involve students choosing between time-based and IDR repayment plans. They find that: being shown information about the incomes of recent college graduates causes students to decrease what they expect their income to be, that being shown that information did not change student's choice of repayment plans, and that students are statistically significantly more likely to select the repayment plan framed as the default plan. In Brownstein (2020) I field a small online survey to students at MSU where they choose either an IDR or non-IDR student loan repayment plan. Although many of my results are not statistically significant, I find that students were more likely to choose the IDR plan when: the amount of income not considered when calculating payments is lower, the percent of non-exempt

income determining payment is lower, and the number of years until loan forgiveness is lower. Muller and Yannelis (2019b) study a field experiment where borrowers are randomly sent or not sent pre-populated applications to enroll in an IDR plan. They find that individuals who receive the applications have much higher enrollment in IDR plans, lower loan payments, and a lower probability of failing to make a required loan payment.

The method for eliciting distributional income expectations used in this study comes from Delavande and Rohwedder (2008). They find that, compared to eliciting expectations by asking for points on the cumulative distribution function, eliciting expectations by asking respondents to place balls in bins representing ranges of the probability distribution leads to a statistically significantly higher percentage of respondents with valid probability distributions. Delavande, Giné, and McDenzie (2011) find that using this method to elicit income expectations in developing countries provides reasonable responses that are predictive of future economic behavior. Orr (2020) uses this method to elicit the subjective expectations of college students, including questions about expected GPA conditional on a certain amount of studying, and questions about income conditional on graduating with a certain GPA.

IV. Description of Survey

This paper analyzes data from a web survey of MSU college seniors²⁰. MSU's Office of the Registrar sent out emails that I wrote on October 19th, October 22nd, and October 25th, 2021. The emails described the survey and had a URL which could be used to take the survey. The emails also informed students that if they completed the survey, they could be sent \$10 using either Venmo or Paypal. The 3 emails were sent to the same 7,000 students. The survey was closed on October 27th, 2021. Screenshots of the emails are available upon request. Before any data

²⁰ MSU's Office of Financial Aid defines a senior as an undergraduate student who has completed at least 88 credits.

was analyzed, incomplete survey responses and any response after the first response by the same person were removed²¹. After that 1,581 responses were left. The survey has a response rate of 22.6%. The median time it took students in the sample to complete the survey is 9 minutes and 56 seconds.

Survey respondents are asked about their income expectations in the form of a statistically valid probability distribution. The method of eliciting this distribution comes from Delavande and Rohwedder (2008). Survey respondents allocate 10 balls to the following income ranges: \$0 - \$30,000, \$30,000 - 60,000, \$60,000 - \$90,000, \$90,000 - \$120,000, and greater than \$120,000. Each ball they allocate to an income range represents a 10-percentage point probability that they expect to earn an annual income in that range. Survey respondents are asked about what income they expect to receive 5 years after graduating with an undergraduate degree from MSU²². Survey respondents are asked not to count any time in graduate or professional school as part of those 5 years²³.

Each time after they are asked for their income expectations, survey respondents are asked to choose between two different repayment plans. They are asked to assume they have graduated from MSU with \$30,000 in student loan debt, and the debt has an interest rate of 5%. Repayment Plan 1 is an IDR plan like the widely available Revised Pay as You Earn Plan. Repayment Plan 2 is a time-based repayment plan like the Standard Repayment Plan. Information about the repayment plans is shown in three tables. The first table describes the two repayment plans. The other tables have estimates of minimum monthly payments,

²¹ In cases where 2 or more responses had the same Venmo account name or email for Paypal, all responses except for the response with the earliest recorded date were deleted. 4 completed responses had neither a Venmo account nor an email for Paypal and therefore could not be checked against other responses.

²² Arcidiacono, Hotz, Maurel, and Monamo (2020) survey students at Duke University about their major and occupation specific earnings expectations in 2009. In 2015 they collect data on the survey respondent's actual earnings. They find that a student's earnings expectations are informative about future earnings and that students sorted into occupations based on expected earnings.

²³ This was for two reasons. First, individuals in graduate or professional school have an unusually low income given their level of education. Second, borrowers who are in graduate or professional school can get a deferment and temporarily lower their required loan payment to \$0.

estimated length of time making payments, and total amount paid over the course of the loan. These estimates are given for the two repayment plans for starting post-college incomes of between \$10,000 and \$90,000 in \$10,000 increments²⁴.

After being asked about their income expectations and choice of repayment plan for the first time, survey respondents are randomly shown one of the two information treatments described below.

One information treatment contains information on the average yearly incomes of individuals in the U.S. with a college degree²⁵. I refer to this treatment as the All-Graduates Income Treatment. This statistic is calculated using the American Community Survey 2014 – 2018 IPUMS file (Ruggles, Flood, Goeken, Grover, Meyer, Pacas, and Sobek 2020). This information is intended to be a placebo treatment in that it would not change a survey respondent's income expectations. I expected that students would think that information about the incomes of college graduates of all ages is too general to affect their earnings expectations. The purpose of including a treatment like this is to deal with issues related to the Hawthorne effect and to have a control group without letting survey respondents know that they are in the control group.

The other treatment shows survey respondents the median yearly earnings of MSU graduates with majors similar to the respondent's primary major. I call this treatment the Major Specific Income Treatment. The median earnings data is from the U.S. Department of Education's College Scorecard²⁶. The College Scorecard has data on median earnings for students of either a single major or a group of related majors. The median earnings statistic that a student who received the Major

²⁴ See Appendix A for screenshots from survey. Total amount paid and length of time making payments are calculated assuming simple daily interest and income increasing at 5% on January 1st of each year. Additional details about those calculations are available upon request.

²⁵ I calculated the average income to be \$53,268.

²⁶ The data was taken from the following URL in October of 2020: <https://collegescorecard.ed.gov/school/fields/?171100-Michigan-State-University>. The data matched the median salary 1 year after graduation (EARN_MDN_HI_1YR) for 2015-2016 and 2016-2017 classes that can be downloaded from the College Scorecard's data website (<https://collegescorecard.ed.gov/data/>).

Specific Income Treatment sees is based on the survey respondent's reported primary major. The statistics shown are for median earnings during the first year after students have graduated from MSU. Only students who got federal financial aid are included in the sample to calculate the medians.

My hypothesis is that the major specific earnings data would increase the probability students expected to earn a low income, and that this would cause them to be more likely to choose the IDR plan. Cox, Kreisman, and Dynarski (2020) study student loan repayment plan choice by randomly providing or not providing students with information related to their future income. In that study, about half of college students who participated in a laboratory experiment are provided information on the distribution of earnings of 24-year-old bachelor's degree holders. Those who see the information expect themselves and their peers to earn statistically significantly less than experiment participants who are not provided with that information. Based on this, I expect that providing students with information about the earnings of recent BA holders would shift their expected income distribution to center around lower incomes. This in turn would increase students' subjective probability that they would earn a low income. In my survey, given the hypothetical borrowing amount, available plans, and interest rate, borrowers whose annual income is less than \$58,184 would have lower required monthly payments on the IDR plan described in the survey than if they were on the non-IDR plan described in the survey.

After being shown one of the treatments, respondents are then again asked the same questions related to income expectations and repayment plan choice. Then survey respondents are asked four questions to test their understanding of the two repayment plans. See Appendix A for screenshots of these questions. The survey ends with a series of questions related to the survey respondent's demographics and their college financial aid. This section includes questions about the survey

respondent's gender, race, and age. The survey respondents are also asked how much student loan debt they have²⁷.

V. Empirical Framework

The goal of this research project is to use a randomized information treatment to create exogenous variation in low-income expectations between two groups of students. I then want to see if the group which believes they have a higher probability of earning a low income is more likely to choose the IDR plan.

To study how low-income expectations and repayment plan choice are affected by the treatments, I use the following estimating equations:

$$(1) Outcome_i = \beta_0 + \beta_1 sawMajorSpecificIncome_i + \beta X_i + \epsilon_i$$

$$(2) Outcome_{it} = \beta_0 + \beta_1 sawMajorSpecificIncome_i + \beta_2 AfterTreatment_t + \beta_3 sawMajorSpecificIncome_i * AfterTreatment_t + \epsilon_{it}$$

i indexes the survey respondent. $Outcome_i$ is two different variables. One variable is the subjective probability a student believes they would earn a low-income. For my main analysis, I define earning a low income as earning \$0 to \$30,000 a year. A second main variable is an indicator variable that equals 1 if the respondent chooses the IDR plan and equals 0 if the respondent chooses the non-IDR plan. X_i is a vector of covariates. The covariates in the analysis are the same covariates I use in the balance tests. X_i includes indicator variables for having a single major, being female, being white, having a Pell Grant, being a first-generation college student, and having student loans. X_i also includes continuous variables for a student's age and the student's subjective probability that they will attend graduate or professional school in the next 20 years.

²⁷ Past research which compares how much student loan debt students say they have in surveys to university administrative records of student loan debt has found that many students do not correctly report how much student loan debt they have (Akers and Chingos, 2014; Andruska, Hogarth, Fletcher, Robes, and Wohlgemuth 2014). I use data on student loan debt only to categorize survey respondents who do and do not have student loans. Unfortunately, even this categorization likely has measurement error. Andruska, Hogarth, Fletcher, Robes, and Wohlgemuth (2014) find that 62 of 165 students in their study who reported in a survey they had no student loan debt had student loan debt in administrative records.

Equation 1 uses data from the income expectations and repayment plan choice questions that are asked after the survey respondents see one of the treatments. The coefficient of interest in equation 1 is β_1 . β_1 is the average outcome for survey respondents who see the Major Specific Income Treatment minus the average outcome for survey respondents who see the All-Graduates Income Treatment.

Equation 2 uses data from the income expectations and repayment plan choice questions that are asked both before and after the survey respondent has seen information about the income of college graduates. t indexes when the outcome is measured in the survey. Either $t = 0$ when the outcome is measured before the income information is shown or $t = 1$ after the income information is shown. $AfterTreatment_t$ is an indicator variable for the outcome being recorded after the survey respondent has seen the income information. The coefficient of interest in equation 2 is β_3 . Equation 2 improves on equation 1 by controlling for pre-treatment differences in the outcome variable.

The following equation is used to see if the difference in the effect of the treatments on outcomes are statistically significantly different for different subgroups.

$$(3) Outcome_{ist} = \beta_0 + \beta_1 sawMajorSpecificIncome_i + \beta_2 SubgroupMember_s + \beta_3 AfterTreatment_t + \beta_4 sawMajorSpecificIncome_i * SubgroupMember_s + \beta_5 sawMajorSpecificIncome_i * AfterTreatment_t + \beta_6 SubgroupMember_s * AfterTreatment_t + \beta_7 sawMajorSpecificIncome_i * SubgroupMember_s * AfterTreatment_t + \epsilon_{ist}$$

Equation 3 regresses outcomes for respondent i in subgroup s measured at time t . $SubgroupMember_s$ is an indicator variable for being a member of a subgroup such as having student loans or having a low-income major. Equation 3 has three sources of variation: the variation in outcome by treatment, the variation in outcome

by subgroup, and the variation in the outcome before the treatment and after the treatment. The coefficient of interest in Equation 3 is β_7 . β_7 can be thought of as how much the effect of the Major Specific Income Treatment is different for survey respondents who are and are not members of the subgroup, controlling for pre-treatment differences in the outcome by subgroup and treatment.

VI. Results

A. Analysis Sample

Before any data is analyzed, incomplete survey responses and any response after the first response by the same person are removed²⁸. After that, 1,581 responses are left. The survey has a response rate of 22.6%. 38 international students are removed from the sample because only U.S. citizens are eligible for student loans from the U.S. government. 147 additional students are dropped because of missing income information²⁹. This leaves an analysis sample of 1,396 completed responses.

B. Summary Statistics

Appendix B contains summary statistics for the analysis sample. The sample contains individuals with 95 different primary majors. The 5 majors with the highest number of individuals in the sample are: Human Biology (107 respondents), Psychology (68 respondents), Finance (66 respondents), Neuroscience (62 respondents), and Kinesiology (59 respondents). 87% of respondents reported having only 1 major when they took the survey. Each MSU major is matched to a description of a major or group of majors in the College Scorecard to determine what income would be shown if the respondent received the Major Specific Income

²⁸ In cases where 2 or more responses had the same Venmo account name or email for Paypal, all responses except for the response with the earliest recorded date were deleted. 4 completed responses had neither a Venmo account nor an email for Paypal and therefore could not be checked against other responses.

²⁹ These individuals are removed from the analysis because, based on their primary major, if they are selected to receive the Major Specific Income Treatment, they would see a median income of Data Not Available. For many of these majors, there would be a major or group of majors in the College Scorecard dataset that was like a particular MSU major. However, in the College Scorecard dataset the median earnings for MSU graduates for the major or group of majors was listed as unavailable. A list of which majors were or were not in the analysis sample is available upon request.

Treatment. Survey respondents were matched to 61 College Scorecard major descriptions, with some College Scorecard major descriptions being matched to more than one MSU major. The top 5 College Scorecard major descriptions in the data are: Physiology, Pathology, and Related Science (137 respondents), Psychology (68 respondents), Finance and Financial Services Management (66 respondents), Public Relations, Advertising, and Applied Communications (65 respondents), and Business Administration, Management, and Operations (63 respondents). 59% of the sample is female and 81% is white. The average age respondents is 21. 59% of respondents have student loans and 31% have ever had a Pell Grant. 19% of respondents are first-generation college students. Individuals in the sample believed they had an average subjective probability of 65% of attending graduate or professional school in the next 20 years³⁰.

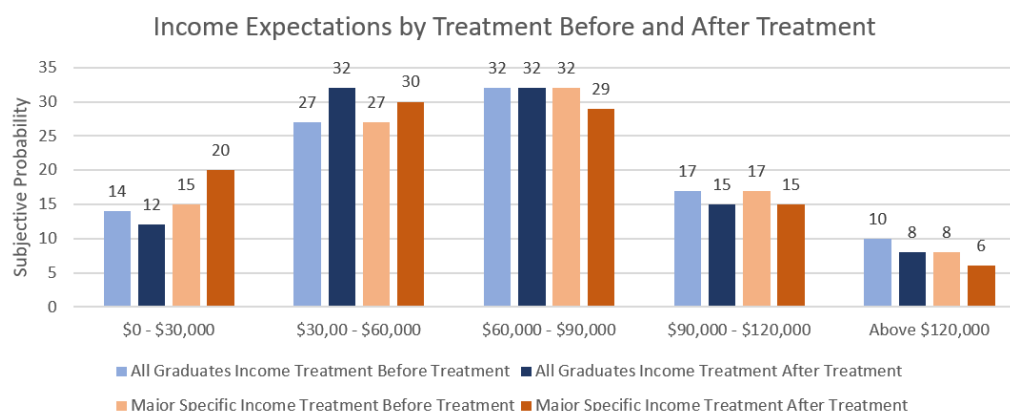
C. Balance Tests

In Appendix C, I test for balance in covariates between survey respondents who saw the All-Graduates Income Treatment and survey respondents who saw the Major Specific Income Treatment. I regress a binary variable for seeing the Major Specific Income Treatment on 8 binary variables related to a survey respondent's demographics and college financial aid. The only coefficient for a covariate that is statistically significant at the 5% level is the coefficient for having a single major. An F-test of joint significance for that regression has a p-value of 0.2376. Therefore, I believe the covariates are balanced across the treatments. Because of that, I interpret coefficient estimates on the coefficients of interest as causal effects of seeing the Major Specific Income Treatment on the outcome compared to seeing the All-Graduates Income Treatment.

³⁰ 27% of individuals were either in or seeking continuing education 6 months after graduating from MSU. See <https://careernetwork.msu.edu/outcomes/> Accessed November 11th, 2021. Using data from the 2007 – 2008 Baccalaureate and Beyond Longitudinal Study, Baum and Steele (2017) estimate that 39% of individuals who graduated with a bachelor's degree in 2007 – 2008 enrolled a graduate degree program within 4 years of graduating from college.

D. Distribution of Income Expectations by Treatment Before and After Treatment

Figure 1 - Income Expectations by Treatment Before and After Treatment



Notes: N = 2,792. Because each respondent gave their income expectations twice, each respondent has 2 observations.

Figure 1 shows the average distribution of income expectations for each treatment before and after seeing the income information. Firstly, this figure shows that both treatments have similar average distributions of income expectations before the income information is shown. For the two treatments, the middle three income ranges have the exact same average subjective probability, and the other two income ranges are different by no more than 2 percentage points.

Second, this figure shows that both treatments cause income expectations to change. Survey respondents who see the All-Graduates Income Treatment believe they have a higher average probability of receiving an income between \$30,000 and \$60,000, and a lower or no different probability of earning an income in other income ranges after they see the income information. The income information causes survey respondents to believe they have an increased probability of receiving an income close to the typical income number they see. In the case of the All-Graduates Income Treatment this income is \$53,268. This is different than my expectation that the information in the All-Graduates Income Treatment would be

too general for it to affect a survey respondent's income expectations. Survey respondents who see the Major Specific Income Treatment increase the probability they believe they would earn between \$0 and \$30,000 and between \$30,000 and \$60,000. Those survey respondents also have a decreased average probability they believe they would earn an income in the other three income ranges. Given that 83% of survey respondents who see the Major Specific Income Treatment see a typical income less than \$60,000, this is consistent with survey respondents responding to income information by increasing the probability they believe they will earn an income close to the income that they see.

Having the All-Graduates Income Treatment change survey respondents' income expectations does not invalidate my research design. So long as the two treatments create exogenous variation in low-income expectations, I can relate differences in low-income expectations, uncorrelated with anything else, to differences in repayment plan choice. However, having the All-Graduates Income Treatment change income expectations means that I do not have evidence for how students would change their income expectations and repayment plan choice if they were simply asked questions about income expectations and repayment plan choice twice.

E. *Effect of Treatment on Low Income Expectations*

Table 1 – Effect of Treatment on Low Income Expectations

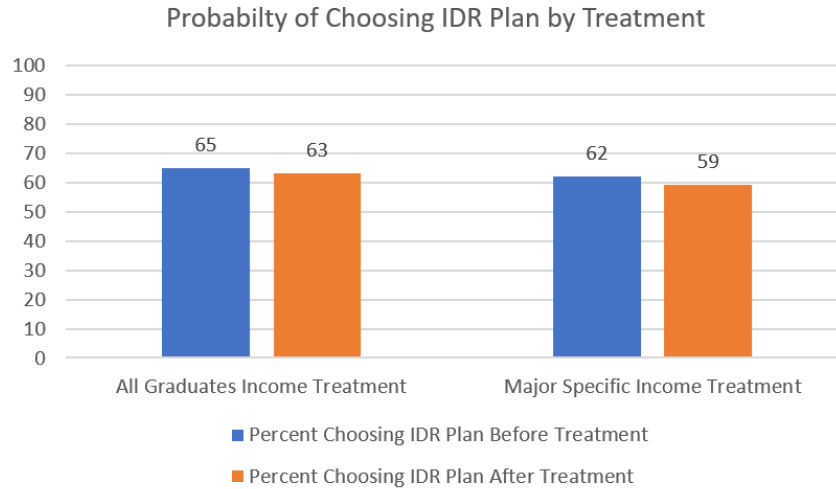
	Percent Chance of Earning a Low Income	Percent Chance of Earning a Low Income
Major Specific Treatment	7.9997*** (1.2381)	8.0647*** (1.2152)
Covariates	N	Y
N	1,396	1,396
	Percent Chance of Earning a Low Income	
Major Specific Treatment * After Treatment	7.0947*** (1.6578)	
Covariates	N	
N	2,792	

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For the regression with Major Specific Treatment * After Treatment each survey respondent has two observations: one observation before the treatment and one after the treatment. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believed they would earn between \$0 and \$30,000 5 years after graduating from MSU regressed on binary variables for the survey respondent seeing the major specific income treatment (Major Specific Treatment) or a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. Standard errors are robust standard errors. Covariates included in the regression are binary variables for the survey respondent: being female, being white, having only 1 major, having a Pell Grant, being a first-generation college student and having student loans and discrete variables for the survey respondent's age and the probability the survey respondent believed they would attend graduate or professional school within 20 years of answering the survey.

Table 1 shows estimates of how seeing the Major Specific Income Treatment changes low-income expectations relative to seeing the All-Graduates Income Treatment. Seeing the Major Specific Income Treatment causes students to believe they had, on average, an 8-percentage point higher probability of earning a low income compared to if they saw the All-Graduates Income Treatment. Controlling for covariates changes the estimate very little consistent with covariates being balanced across treatments. Controlling for pre-treatment differences in low-income expectations reduces the treatment effect to 7-percentage points. In all cases the effect is statistically significant.

F. Effect of Treatment on Repayment Plan Choice

Figure 2 - Plan Choice by Treatment Before and After Treatment



Notes: N = 2,792. Because each respondent chose a repayment plan twice, each respondent has 2 observations.

Table 2 – Effect of Treatment on Low Income Expectations

	Choose IDR Plan	Choose IDR Plan
Major Specific Treatment	-0.0432* (0.0261)	-0.0455* (0.0262)
Covariates	N	Y
N	1,396	1,396
	Choose IDR Plans	
Major Specific Treatment * After Treatment	-0.0187 (0.0367)	
Covariates	N	
N	2,792	

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For the regression with Major Specific Treatment * After Treatment each survey respondent has two observations: one observation before the treatment and one after the treatment. This table shows the results of an indicator variable for the survey respondent choosing the IDR plan regressed on binary variables for the survey respondent seeing the major specific income treatment (Major Specific Treatment) or a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about repayment plan choice coming after the treatment (After Treatment) and an interaction between those variables. Standard errors are robust standard errors. Covariates included in the regression are binary variables for the survey respondent: being female, being white, having only 1 major, having a Pell Grant, being a first-generation college student and having student loans. It also includes as covariates discrete variables for the survey respondent's age and the probability the survey respondent believed they would attend graduate or professional school within 20 years of answering the survey.

Figure 2 shows the percent of the analysis sample who saw each treatment who chose the IDR plan before and after seeing the treatment. I hypothesized that

because the Major Specific Income Treatment would increase the subjective probability that survey respondents believed they would earn a low-income, the Major Specific Income Treatment would cause survey respondents to be more likely to choose the IDR plan. Contrary to my hypothesis, all 3 regressions in Table 2 estimate that the treatment effect of the Major Specific Income Treatment is to decrease the percent of survey respondents who choose the IDR plan. The effect is about -4 percentage points without covariates, -5 percentage points with covariates, and -2 percentage points controlling for pre-treatment differences in plan choice. No coefficient is statistically significant. When controlling for pre-treatment differences in plan choice the 95% confidence interval of the treatment effect is -9 percentage points to 5 percentage points. This is despite that, consistent with my hypothesis, seeing the Major Specific Income Treatment caused students to have a higher average subjective probability of earning a low income compared to seeing the All-Graduates Income Treatment.

G. Robustness Checks

In Appendix C I find a statistically significant difference between the treatments in the proportion of survey respondents who have only one major. This difference may help explain differences in outcome by treatment. In results available upon request, I recreate Figures 1 and 2 and Tables 1 and 2 dropping all students with more than one major. The results are similar with a statistically significant difference in low-income expectations by treatment, but not a statistically significant difference in plan choice by treatment. The magnitude of all coefficients of interest are within 1 percentage point in these tables compared to coefficients of interest for the full sample.

To see if the low-income expectations results are robust to a change in the highest income that is considered low, I replicated the analysis on low-income expectations changing the definition of low-income from earning between \$0 and \$30,000 to earning between \$0 and \$60,000. The results of this analysis are in

Appendix D. Survey respondents who received the Major Specific Income Treatment have a subjective probability of earning between \$0 and \$60,000 after the survey respondents see the income information that is 5 to 6 percentage points higher than survey respondents who saw the All-Graduates Income Treatment. This after treatment difference is statistically significant at the 5% level with or without covariates. Considering pre-treatment differences in income expectations reduces this difference to about 4 percentage points. In the regression using both before and after treatment income expectations, this lower coefficient value and a much larger standard error makes this difference not statistically significant at the 10% level.

H. Heterogeneity by Having Student Loans

Only 41% of survey respondents in the analysis sample report they have student loans³¹. Because only students with student loans must choose a student loan repayment plan, I want to see if the results are similar for respondents with and without student loans. Appendix E shows the results of this analysis.

To see how survey respondents with student loans are different from survey respondents without student loans, I test for statistically significant differences in other covariates for survey respondents with and without student loans using a multivariate regression. The results of this analysis are in Table E1. Survey respondents with student loans are statistically significantly more likely to have a Pell Grant (23-percentage points) and to be a first-generation college student (11-percentage points). Overall, these results indicate that the main difference for students with and without student loans in my sample is that survey respondents with student loans come from families with a lower socio-economic status than those without student loans.

³¹ Using information in the Common Data Set voluntarily reported by Michigan State University, The Institute for College Access & Success concluded that 25% of student debt for college graduates at Michigan State University was non-Federal. See <https://ticas.org/wp-content/uploads/2020/10/Michigan.pdf>. The major source of non-Federal student loan debt is student loans given by private sector financial companies. These companies, as far as I know, do not offer IDR plans. Because I did not ask if a survey respondent's loans were Federal or private, I am unable to know which survey respondents with student loan debt had private student loans.

The treatment effect for seeing the Major Specific Income Treatment for survey respondents with student loans on low-income expectations is a not statistically significant 4.72 percentage points higher than it is for survey respondents without student loans. For survey respondents with student loans, the treatment effect of seeing the Major Specific Income Treatment on earning a low income is a significant 9.09 percentage points. For survey respondents without student loans this effect is a not statistically significant 4.37 percentage points.

The treatment effect for seeing the Major Specific Income Treatment for survey respondents with student loans on the probability of choosing an IDR plan is a not statistically significant -6.06 percentage points different than it is for survey respondents without student loans. For survey respondents with student loans, the treatment effect of seeing the Major Specific Income Treatment is a not statistically significant decrease in the probability of choosing an IDR plan by 4.24 percentage points. For survey respondents without student loans this effect is a not statistically significant increase in the probability of choosing an IDR plan of 1.81 percentage points.

I. Heterogeneity by Income of Major

Appendix F contains tables and figures that look at the heterogeneity of results by the income of the survey respondent's major. For this analysis, the survey respondent's major income is equal to the income the respondent would see if they are chosen to receive the Major Specific Income Treatment. This means the major income is the income of MSU graduates with federal financial aid one year after they graduated with a major similar to the survey respondent's primary major. The median major income for the sample is \$37,400. Survey respondents are classified as having a low-income major if their major income is below the sample median major income. Survey respondents are classified as having a high-income major if their major income is equal to or above the sample median major income.

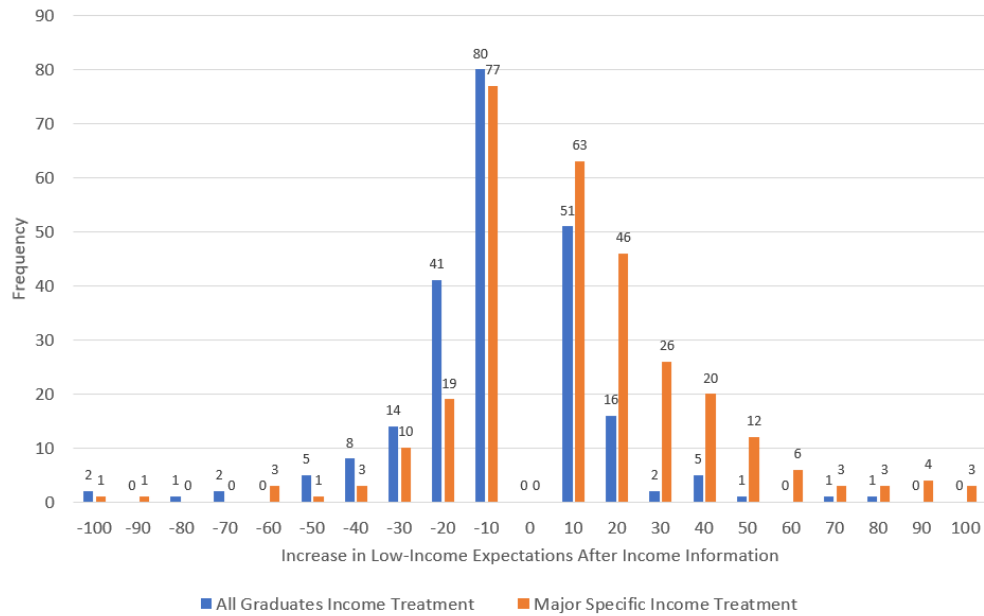
The treatment effect for seeing the Major Specific Income Treatment for survey respondents with a low-income major on low-income expectations is a statistically significant 13.88 percentage points higher than it is for survey respondents with high-income majors. For survey respondents with a low-income major, the treatment effect of seeing the Major Specific Income Treatment on a survey respondent's subjective probability of earning a low income is a statistically significant 13.91 percentage points. For survey respondents with a high-income major this effect is a not statistically significant 0.04 percentage points.

The treatment effect for seeing the Major Specific Income Treatment for survey respondents with a low-income major on the probability of choosing an IDR plan is a not statistically significant 8.05 percentage points different than it is for survey respondents with a high-income major. For survey respondents with a low-income major, the treatment effect of seeing the Major Specific Income Treatment is a not statistically significant increase in the probability of choosing an IDR plan by 2.06 percentage points. For survey respondents with a high income major this effect is a not statistically significant decrease in the probability of choosing an IDR plan of 5.99 percentage points.

J. Change in Income Expectations, Change in Repayment Plan Choice

One reason for the small effect of the Major Specific Income Treatment on repayment plan choice may be that few survey respondents changed their low-income expectations when they see the income information. Overall, 38% of survey respondents changed their low-income expectations after they see the income information. 33% of survey respondents who see the All-Graduates Income Treatment change their low-income expectations. 43% of respondents who see the Major Specific Income Treatment change their low-income expectations.

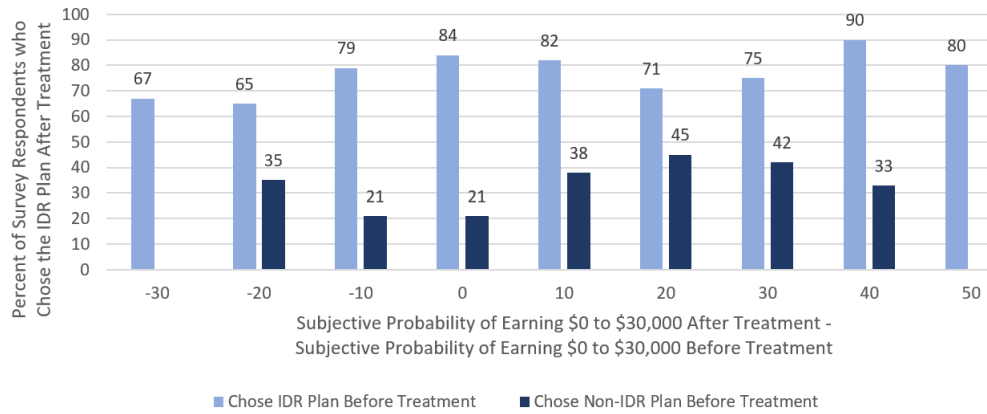
Figure 3 - Increase in Low-Income Expectations After Treatment by Treatment



Notes: N = 531. 865 respondents who did not change their Low-Income Expectations after receiving the income information are not shown in the figure to make it easier to read.

Figure 3 shows the distribution in the increase of low-income expectations after the income information separately for the All-Graduates Income Treatment and the Major Specific Income Treatment. Survey respondents who did not change their low-income expectations were removed to make the figure easier to see. Survey respondents who saw the Major Specific Income Treatment are more likely to increase their subjective probability of earning a low income while survey respondents who saw the All-Graduates Income Treatment are more likely to decrease their subjective probability of earning a low-income.

Figure 4: Plan Choice After Treatment by Change in Low-Income Expectations



Notes: N = 1,337

Figure 4 shows the relationship between the change in survey respondents' low-income expectations and the probability respondents choose the IDR plan after the treatment³². If low-income expectations were strongly related to repayment plan choice, I would expect the probability of choosing the IDR plan after the treatment to be higher for survey respondents who had a higher increase in their probability of earning a low income after the treatment. In that case, the bars would get higher as you moved to the right on the graph. Visually there is no large consistent increase or decrease in the height of the bars as you move to the right along the graph. This is consistent with low-income expectations having little effect on repayment plan choice.

³² If the number of survey respondents who changed their low-income expectations by a certain number of percentage points conditional on plan choice, like survey respondents who initially chose the IDR plan and whose low-income expectations decreased by 40 percentage points, is less than 10, then those changes in the probability of earning a low income are not shown.

Table 3 - Statistical Test of Relationship Between Change in Low Income Expectations and Change in Plan Choice

	IDR to IDR	IDR to non-IDR	Non-IDR to IDR	Non-IDR to non-IDR	Choose IDR After Treatment	Choose IDR After Treatment
Change in Low Income Expectations	-4.587*10 ⁻⁴ (7.712*10 ⁻⁴)	-6.609*10 ⁻⁴ (5.517*10 ⁻⁴)	1.215*10 ⁻³ ** (5.529*10 ⁻⁴)	-9.55*10 ⁻⁵ (7.419*10 ⁻⁴)	8.833*10 ⁻⁴ (1.019*10 ⁻³)	1.881*10 ⁻³ *
Sample Restrictions	None	None	None	None	Only Choose IDR Plan before Treatment	Choose non-IDR Plan Before Treatment
N	1,396	1,396	1,396	1,396	884	512

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. IDR to IDR is an indicator variable for the survey respondent choosing the IDR plan both before and after the treatment. IDR to non-IDR is an indicator variable for the survey respondent choosing the IDR plan before the treatment and the non-IDR plan after the treatment. Non-IDR to IDR is an indicator variable for the survey respondent choosing the non-IDR plan before the treatment and the IDR plan after the treatment. Non-IDR to non-IDR is an indicator variable for the survey respondent choosing the non-IDR plan both before and after the treatment. Change in Low Income Expectations is the subjective probability (scaled to be between 0 and 100) that the survey respondent believed they would earn between \$0 and \$30,000 5 years after the graduated from MSU after they saw the treatment minus what they believed that subjective probability was before they saw the treatment. Standard errors are robust standard errors.

Table 3 shows the results of regressions of different variables related to plan choice on a survey respondent's change in the probability they believe they would earn a low income. The only regression which has a coefficient that is statistically significant at the 5% level is the regression on a survey respondent switching from preferring the non-IDR plan before seeing a treatment to the IDR plan after seeing a treatment. Based on that regression, a 10-percentage point increase in the probability the survey respondent believed they would earn a low income is associated with an increased probability of switching from the non-IDR plan to the IDR plan of 1.2 percentage points.

The results in Figure 4 and Table 3 show that, for the most part, how a survey respondent changed their expectations of earning a low-income after seeing the treatment is not related to if they changed which repayment plan they preferred after seeing the treatment.

VII. Discussion

Contrary to my hypothesis, I find that the Major Specific Income Treatment did not cause survey respondents to be statistically significantly more likely to choose the IDR plan. This is despite the Major Specific Income Treatment causing survey respondents to have a statistically significantly higher subjective probability of

earning a low-income. When controlling for pre-treatment differences in covariates, the 95% confidence interval of the treatment effect of the Major Specific Income Treatment on choosing an IDR plan is -9 percentage points to 5 percentage points. This means it is highly unlikely the treatment effect of the Major Specific Income Treatment is above 5 percentage points. I conclude the survey respondents in my sample are not choosing a student loan repayment plan based on minimizing their required payments in the event they have a low-income.

One possible explanation for the results is that survey respondents are worried about the costs of making low payments on their student loans. In general, if a borrower makes a smaller monthly payment on their student loans, they will have to pay more interest over the life of the loan. A lower monthly loan payment will also cause a respondent to take longer to pay off the loan. Survey respondents are shown tables with information about starting monthly payments, estimated total amount of money paid on the loan, and estimated total time making payments for different starting levels of income for both the non-IDR and IDR plan. This information is on the page where students are asked to choose either an IDR or non-IDR plan. These estimates assume a survey respondent's income increases by 5% at the start of every year³³.

Whether or not a survey respondent would pay more on their student loans if they made lower payments depends on how much of their loans are forgiven. I estimate that if the survey respondent made either \$30,000, \$40,000, or \$50,000 when they began making payments on their loans, they their total payments would be higher if they were on the IDR plan³⁴. However, because the IDR plan forgives any remaining loan balance after 20 years of payments, a student who is on the IDR plan described in the survey and had \$30,000 of student loan debt when they

³³ The Department of Education's assumes borrower's incomes increase by 5% per year when they estimate future student loan payments on their exit counseling website.

³⁴ This assumes the survey respondent starts out with \$30,000 in student loan debt and makes the minimum required payment every month after they started paying back their loans while they have a positive loan balance.

graduated MSU, might have lower total loan payments on the IDR plan, even if their required monthly payments are generally less than they would be if they were on the non-IDR plan. I estimate this happens if a survey respondent's income is either \$10,000 or \$20,000 when they begin making payments. Even if survey respondents understood the benefits of being on an IDR plan in terms of total payments when their starting income is \$10,000 or \$20,000, the cost of increased total payments when their starting income is \$30,000, \$40,000, or \$50,000 may have made them less likely to change from the non-IDR plan to the IDR plan when they believed they had a higher probability of earning a low-income.

To test if concerns about having higher total payments might explain the lack of an effect of the Major Specific Income Treatment on repayment plan choice, in Appendix G I analyze the effect of the Major Specific Income Treatment on a survey respondent's subjective probability of earning between \$30,000 and \$60,000 a year. It is in that income range that I estimate a survey respondent would have higher total payments on the IDR plan than the non-IDR plan. If the Major Specific Income Treatment causes survey respondents to believe they have a higher probability of earning \$30,000 to \$60,000 a year, the possible additional costs to them in terms of total payments would help explain why the Major Specific Income Treatment did not increase the probability respondents chose the IDR plan. Using data from before and after the treatments, the treatment effect of the Major Specific Income Treatment on that probability is a not statistically significant -2.98 percentage points. For all specifications in Appendix G the treatment effect is negative. This result is not consistent with higher total payments explaining the lack of an effect of the treatment on plan choice.

Another thing survey respondents may have been concerned about is how long they would have student loans. Even if they earned less than \$20,000, and therefore would not be required to make payments if they were on the IDR plan, they would still have student loan debt. If having student loan debt imposes a mental cost on

individuals no matter the level of payments, then even respondents who earned a low-income would have a reason to make higher payments so they could be debt free sooner. This may have discouraged survey respondents who had a higher probability of earning a low-income from choosing the IDR plan. If a borrower's starting salary is \$10,000, \$20,000, \$30,000, or \$40,000, the table in the survey has an estimated time making payments for the IDR plan of longer than the 10 years of payments on the non-IDR plan. There are both costs and benefits to having a low-income while being on the IDR plan. This may help explain why causing survey respondents to have a higher subjective probability of earning a low-income did not coincide with being more likely to choose an IDR plan.

Finally, it may be the case that survey respondents did not respond to having a greater subjective probability of earning a low-income by being more likely to choose the IDR plan because they did not understand the differences between the IDR plan and the non-IDR plan. Survey respondents only correctly answer on average 1.8 of the 4 questions testing their understanding of the two repayment plans. This is not statistically significantly different by treatment. Survey respondents who see the Major Specific Income Treatment correctly answer 0.11 fewer questions than survey respondents who see the All-Graduates Income Treatment³⁵. Neither is there a statistically significant difference in the number of correct responses between survey respondents with and without student loans³⁶. It may be the case that if survey respondents better understood the plans, they would have responded to having a greater subjective probability of earning a low income in the way I predicted.

I test if the level of repayment plan understanding is related to the treatment effect of the Major Specific Income Treatment in Appendix H. To do this, I

³⁵ P-Value 0.155

³⁶ Survey respondents without student loans answer on average 1.86 questions correct while survey respondents with student loans answer on average 1.82 questions correct. The P-value for the difference is 0.569.

compare the treatment effect for survey respondents who answer 0 or 1 of the 4 questions correct, I label those survey respondents having low plan understanding, to the treatment effect for students who answer 2 or more questions correct, I label those survey respondents as having a high plan understanding. For both survey respondents with high and low plan understanding, the Major Specific Income Treatment causes survey respondents to believe that they have a statistically significantly higher probability of earning a low income. However, for neither group did the Major Specific Income treatment statistically significantly increase the probability of choosing the IDR plan. The treatment effects for the two groups were not statistically significantly different for either low-income expectations or repayment plan choice. Based on this, I do not think low understanding of the repayment plans explains why the Major Specific Income Treatment did not statistically significantly increase the probability a survey respondent chose the IDR plan.

VIII. Conclusion

In this paper I test the hypothesis that a student's subjective probability of earning a low income is a causal factor in if they prefer an IDR or non-IDR student loan repayment plan. I predict that students who had an exogenously higher subjective probability of earning a low income would be more likely to choose an IDR plan. I test that using data from a web survey emailed to undergraduate seniors at MSU. The survey randomizes the type of information about post-college incomes survey respondents are shown to create two groups of survey respondents with exogenously different probabilities of earning a low income. I find seeing the Major Specific Income Treatment causes survey respondents to believe they have a statistically significantly higher probability of earning a low-income compared to survey respondents who see the All-Graduates Income Treatment. Despite this, survey respondents seeing Major Specific Income Treatment do not cause respondents to have a statistically significantly different probability of choosing the

IDR plan then survey respondents who see the All-Graduates Income Treatment. This pattern is similar looking at a variety of sub samples such as survey respondents with and without student loans, survey respondents with lower and higher earning majors, and survey respondents with lower and higher understanding of the repayment plans.

I conclude that changing a student loan borrower's expectation of earning a low income will not statistically significantly change their repayment plan choice. Attempts to increase take-up of IDR plans may have more success focusing on other changes to student loan repayment plan choice such as emphasizing the benefits of IDR plans (Abraham, Filiz-Ozbay, Ozbay, and Turner, 2020) or making an IDR plan the default repayment plan choice (Cox, Kreisman, and Dynarski, 2020).

Future research could explore what borrowers in general, and students in particular, most care about when paying back their loans. Do they care about minimizing required payments, the total amount their loans cost, how long they have any debt, or some combination of the above? How do borrowers in general, and students in particular, balance the tradeoffs between lower monthly payments and increased amount of time having debt? If students had more choices related to how they repaid their student loans, such as having more control over the length of time they had to pay back their student loans on the non-IDR plan, or how payments were calculated as a function of their annual income on an IDR plan, how would they design their repayment plan?

A second line of future research to expand on this research might be digging deeper into student's expectations of their futures. How do students expect their income to change over time? How do students expect their incomes to change if they attend graduate or professional school? How accurate are student's beliefs related to how much they will earn and how likely they are to attend graduate or professional school? What do students think they will be doing if they earn different ranges of income?

REFERENCES

- Abraham, Katharine G., Emel Filiz-Ozbay, Erkut Y. Ozbay, and Lesley J. Turner.** 2020. "Framing Effects, Earnings Expectations, and the Design of Student Loan Repayment Schemes." *Journal of Public Economics*. Vol. 183
- Akers, Elizabeth J., and Matthew M. Chingos.** 2014. "Are College Students Borrowing Blindly?" Brown Center on Education Policy, Brookings Institution. https://www.brookings.edu/wp-content/uploads/2016/06/are-college-students-borrowing-blindly_dec-2014.pdf
- Allcott, Hunt, and Dmitry Taubinsky.** 2015. "Evaluating Behaviorally Motivated Policy: Experimental Evidence from the Lightbulb Market." *American Economic Review*. Vol. 105(8): 2501 – 2538
- Andruska, Emily A., Jeanne M. Hogarth, Cynthia N. Fletcher, Gregory R. Forbes, and Darin R. Wohlgemuth.** 2014. "Do You Know What You Owe? Student' Understanding of Their Student Loans." *Journal of Student Financial Aid*. 44(2): 125 - 148
- Arcidiancono, Peter V., Joseph Hotz, Arnaud Maurel, and Teresa Romano.** 2020. "Ex Ante Returns and Occupational Choice." *Journal of Political Economy*. 128(12): 4475 - 4522
- Baker, Rachel, Eric Bettinger, Brian Jacob, and Ioana Marinescu.** 2018. "The Effect of Labor Market Information on Community College Students' Major Choice." *Economics of Education Review*. 65: 18 – 30
- Baum, Sandy, and Patricia Steele.** 2017. "Who Goes to Graduate School and Who Succeeds?" Urban Institute Report. https://www.urban.org/sites/default/files/publication/86981/who_goes_to_graduate_school_and_who_succeeds_1.pdf
- Betts, Julian R.** 1996. "What Do Students Know about Wages? Evidence from a Survey of Undergraduates." *Journal of Human Resources*. 31(1):27 - 56
- Bleemer, Zachary, and Basit Zafar.** 2018. "Intended College Attendance: Evidence from an Experiment on College Returns and Costs." *Journal of Public Economics*. 157: 184 – 211
- Brownstein, Joshua.** 2020. "A Pilot Survey on Student Loan Repayment Plan Choice." Unpublished. <https://tinyurl.com/Brownstein2020>
- Chapman, Bruce, and Lorraine Dearden.** 2017. "Conceptual and Empirical Issues for Alternative Student Loan Designs: The Significance of Laon Repayment Burdens for the United States", *Annals of the American Academy of Political and Social Science*. 671(1): 249-268
- Conlon, John J.** 2021 "Major Malfunction: A Field Experiment Correcting Undergraduates' Beliefs about Salaries." *Journal of Human Resources*. Vol. 56(3): 922 - 939

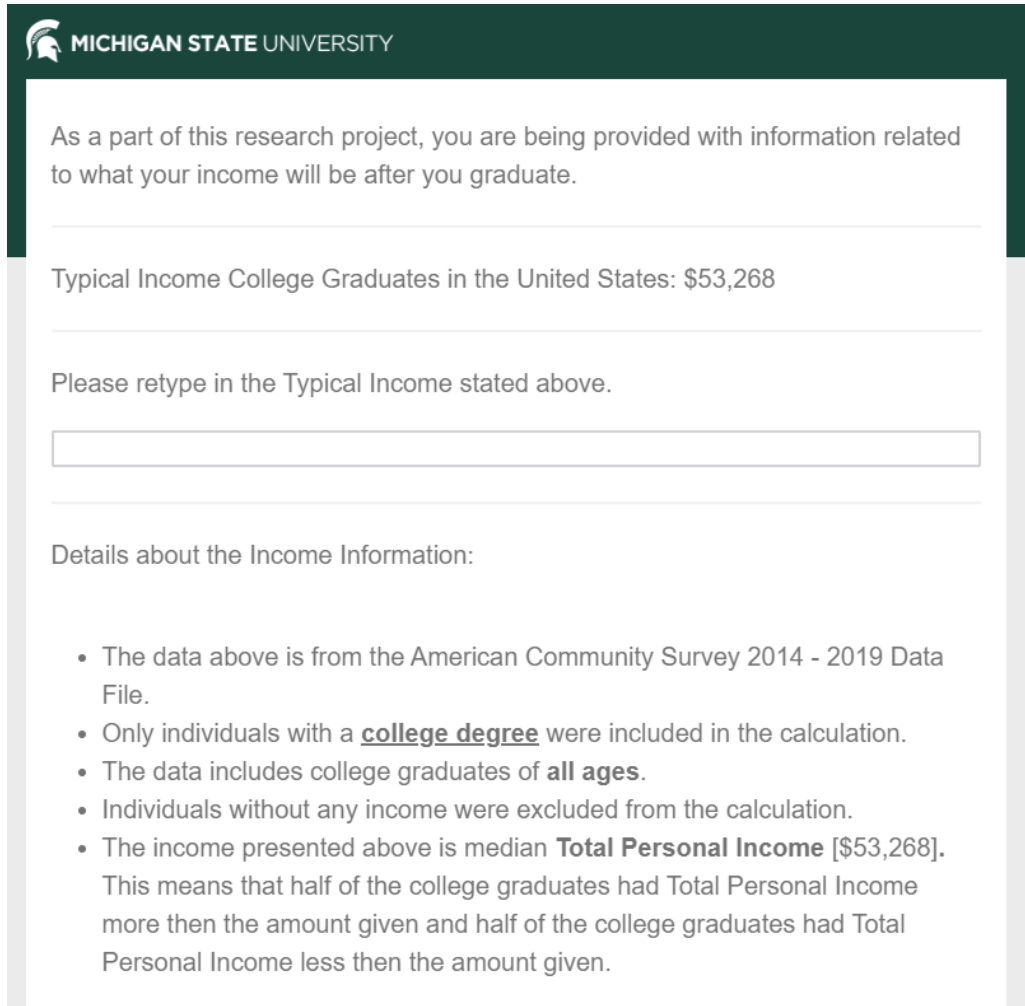
- Cox, James C., Daniel Kreisman and Susan Dynarski.** 2020. “Designed to Fail: Effects of the Default Option and Information Complexity on Student Loan Repayment”. *Journal of Public Economics*. Vol 192
- Darolia, Rajeev, and Casandra Harper.** 2018. “Information Use and Attention Deferment in College Student Loan Decisions: Evidence From a Debt Letter Experiment.” *Educational Evaluation and Policy Analysis*. 40(1): 129 – 150
- Delavande, Adeline, Xavier Giné, and David McDenzie.** 2011. “Measuring Subjective Expectations in Developing Countries: A Critical Review and New Evidence.” *Journal of Development Economics*. 94(2): 151 – 163
- Hastings, Justine, Christopher A. Neilson, and Seth D. Zimmerman.** 2018. “The Effects of Earnings Disclosure on College Enrollment Decisions.” NBER Working Paper 21300. <https://www.nber.org/papers/w21300>
- Herbst, Daniel.** 2023. “The Impact of Income-Driven Repayment on Student Borrower Outcomes.” *American Economic Journal: Applied Economics*. Vol. 15(1): 1 - 25
- Hurwitz, Michael, and Jonathan Smith.** 2018. “Student Responsiveness to Earnings Data in the College Scorecard.” *Economic Inquiry*. 56(2): 1220 – 1243
- Irby, Latoya.** 2019. “What Are the 3 Major Credit Reporting Agencies?” <https://www.thebalance.com/who-are-the-three-major-credit-bureaus-960416> (Accessed February 5, 2020).
- Irby, Latoya.** 2020. “Understanding Your Credit Report and the FCRA” <https://www.thebalance.com/what-you-should-know-about-the-fcra-960639> (Accessed February 5, 2020).
- Kirkham, Elyssa.** 2020. “How Much Money Can I Take Out In Student Loans?”. *Student Loan Hero*. January 14. <https://studentloanhero.com/featured/how-much-student-loans-can-i-get/>
- Klepfer, Kasey, Chris Ferandez, Carla Fletcher, and Jeff Webster.** 2015. “Informed or Overwhelmed? A Legislative History of Student Loan Counseling with a Literature Review on the Efficacy of Loan Counseling”. *TG Research and Analytical Services*. <https://files.eric.ed.gov/fulltext/ED579985.pdf>. (Accessed February 10, 2020)
- Lacy, T. Austin, Johnathan G. Conzelmann, and Nichole D. Smith.** 2018. “Federal Income-Driven Repayment Plans and Short-Term Student Loan Outcomes.” *Educational Researcher*. Vol. 47(4) pp. 255 - 258
- Lane, Ryan.** 2020. “Can You Change Your Student Loan Repayment Plan?”. *Nerdwallet*. January 3. <https://www.nerdwallet.com/blog/loans/student-loans/how-much-student-loans-cost/>

- Ma, Jennifer and Matea Pender.** 2021. “Trends in College Pricing and Student Aid 2021.” College Board. <https://research.collegeboard.org/pdf/trends-college-pricing-student-aid-2021.pdf> (Accessed February 2nd, 2022)
- Mueller, M. Holger, and Constantine Yannelis.** 2019a. “Students in Distress: Home Prices and Student Loan Default in the Great Recession.” *Journal of Financial Economics* 131(1):1-19
- Muller, M. Holger, and Constantine Yannelis.** 2019b. “Reducing Barriers to Enrollment in Federal Student Loan Repayment Plans: Evidence from the Navient Field Experiment.” Unpublished. <https://cmepg.gmu.edu/wp-content/uploads/2019/08/MuellerYannelis.pdf>
- Student Borrower Protection Center.** 2021. “Education Department’s Decades-Old Debt Trap: How the Mismanagement of Income-Driven Repayment Locked Millions in Debt.” National Consumer Law Center. <https://www.nclc.org/uncategorized/issue-brief-education-departments-decades-old-debt-trap-how-the-mismanagement-of-income-driven-repayment-locked-millions-in-debt.html> (Accessed February 17th, 2022)
- Office of the Assistant Secretary for Planning and Evaluation.** 2020. “Poverty Guidelines”. *U.S. Department of Health and Human Services*. January 8. <https://aspe.hhs.gov/poverty-guidelines>
- Orr, Cody.** 2020. “Clocking Into Work and Out of Class: How College Students Make Their Credit Hour Enrollment and Financing Decisions.” Unpublished. <https://orrcody.github.io/research/drafts/into-work-out-of-class.pdf>
- Peter G. Peterson Foundation.** 2021. “10 Key Facts About Student Debt in the United States.” <https://www.pgpf.org/blog/2021/05/10-key-facts-about-student-debt-in-the-united-states> (Accessed February 2nd, 2022)
- Ruggles, Steven, Sarah Flood, Ronald Goeken, Josiah Grover, Erin Meyer, Jose Pacas and Matthew Sobek.** 2020. “IPUMS USA: Version 10.0.” Minneapolis, MN. <https://doi.org/10.18128/D010.V10.0>
- Scott-Clayton, Judith.** 2019. “The Looming Student Loan Default Crisis is Worse than we Thought. Brookings Institution”. Evidence Speaks Reports, Vol 2, #34. <https://www.brookings.edu/wp-content/uploads/2018/01/scott-clayton-report.pdf>, (Accessed December 10, 2019)
- U.S. Government Accountability Office.** 2015. “Federal student loans: education could do more to help ensure borrowers are aware of repayment and forgiveness options”. GAO-15-663.
- Whitsett, Healey, and Rory O’Sullivan.** 2012. “Lost Without a Map: A Survey about Students’ Experiences Navigating the Financial Aid Process.” *NERA Economic Consulting*. https://www.nera.com/content/dam/nera/publications/archive2/PUB_Student_Loan_Borrowers_1012.pdf. (Accessed February 11, 2020)

- Wiswall, Matthew J., and Basit Zafar.** 2015a. "How Do College Students Respond to Public Information about Earnings?" *Journal of Human Capital*. 9(2): 117 – 169
- Wiswall, Matthew J., and Basit Zafar.** 2015b. "Determinants of College Major Choices: Identification from an Information Experiment." *Review of Economic Studies*. 82(2): 791 - 824
- Woo, Jennie, H., Alexander H. Bentz, Stephen Lew, Erin Dunlop Velez, and Nichole Smith.** 2017. "Repayment of Student Loan as of 2015 Among 1995-96 and 2003-04 First-Time Beginning Students" National Center for Education Statistics. <https://nces.ed.gov/pubs2018/2018410.pdf>

Appendix A: Survey Screenshots³⁷

Figure A1 - All Graduates Income Treatment



The screenshot shows a survey interface for Michigan State University. At the top is the university's logo and name. Below this, a message states: "As a part of this research project, you are being provided with information related to what your income will be after you graduate." This is followed by a horizontal line and the text: "Typical Income College Graduates in the United States: \$53,268". Another horizontal line leads to the instruction: "Please retype in the Typical Income stated above." Below this is a text input field. A final horizontal line precedes the section "Details about the Income Information:", which contains a bulleted list of five points regarding the data source (American Community Survey 2014-2019), inclusion criteria (college degree, all ages), exclusions (no income), and the meaning of the median income figure (\$53,268).

MICHIGAN STATE UNIVERSITY

As a part of this research project, you are being provided with information related to what your income will be after you graduate.

Typical Income College Graduates in the United States: \$53,268

Please retype in the Typical Income stated above.


Details about the Income Information:

- The data above is from the American Community Survey 2014 - 2019 Data File.
- Only individuals with a **college degree** were included in the calculation.
- The data includes college graduates of **all ages**.
- Individuals without any income were excluded from the calculation.
- The income presented above is median **Total Personal Income** [\$53,268]. This means that half of the college graduates had Total Personal Income more then the amount given and half of the college graduates had Total Personal Income less then the amount given.

If the survey respondent saw the All-Graduates Income Treatment, they were not allowed to continue to the next page of the survey until they typed 53268 into the textbox. This is done to ensure that survey respondents processed the information they saw.

³⁷ Screenshots for the rest of survey available upon request.

Figure A2 - Major Specific Income Treatment Primary Major Agribusiness Management

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As a part of this research project, you are being provided with information related to what you will earn after you graduate.

Typical Earnings for Agricultural Business and Management **MSU** Graduates:
\$44,300

Please type in the Typical Earnings stated above. If you see "Data Not Available", please type in 0.

Details about the Earnings Information:

- The data above is from the U.S. Department of Education's College Scorecard.
- Only individuals who were **MSU graduates** during the 2015-2016 or 2016 - 2017 academic years were included in the calculation.
- The data is for the **first** year of earnings after graduation.
- The only students included in the calculation were students with federal financial aid.
- Individuals without any earnings were excluded from the calculation.
- The earnings presented above are **Median Earnings**. This means that half of MSU graduates with federal financial aid in the given major earned more than the amount given and half of MSU graduates with federal financial aid in the given major earned less than the amount given.

The major or group of majors and the income that was shown is based on what the survey respondent indicated their primary major is.

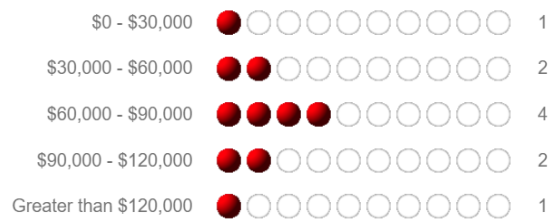
Individuals who saw the above page were not allowed to continue with the survey until they typed 44300 into the textbox. Survey respondents with other majors who saw the Major Specific Income Treatment also had to type in the typical income shown before they could continue to the next page of the survey.

Figure A3 - Question to Elicit Own Income Expectations

What do you expect your income to be 5 years after receiving an undergraduate degree from MSU? Do not count any time in graduate or professional school as part of those 5 years.

Remember, you should place exactly 10 balls.

In order to select 0 in a specific income range, click on the number next to the income range you are interested in changing and type in 0.




This question was asked twice: once before the survey respondent saw the treatment and once after the survey respondent saw the treatment.

The survey respondent was not allowed to continue with the survey until the number of balls they placed in the various income ranges equaled 10.

A valid response with 10 balls in the 5 income ranges is shown in the above picture.

However, survey respondents first see this question with no balls in any of the income ranges.

Figure A4 - Intro to Question Eliciting Repayment Plan Preferences

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Imagine the following scenario.

You have just graduated from MSU **with your current major(s)** with \$30,000 in student loan debt.

You are now required to pick one of the following repayment plans.

The interest rate for both repayment plans is 5%. This is the fee that the lender is charging for having lent you \$30,000.

Payments for both plans are due the first day of each month.

You can pay more than the required payment without penalty.

If you attend graduate or professional school you may pause payments under either plan.

If you are unemployed you may pause payments under either plan.

Figure A5 - Repayment Plan Choice Question Table Describing the Two Repayment Plans

	Minimum Monthly Payment	Loan Forgiveness	Payments Change when Income Changes	Must Pay When Earn Less than \$20,000 a year	Must Pay When in Education	Must Pay When Unemployed
Repayment Plan 1	10% of your annual income above \$20,000 split into 12 monthly payments	Any remaining loan balance after 20 years of payments is forgiven	Yes	No	No	No
Repayment Plan 2	Payments are set so the loan is paid off in 10 years	No loan forgiveness	No	Yes	No	No

In the survey this table appears directly below the information in Figure A4.

Figure A6 - Repayment Plan Choice Question Payment Information Tables

Repayment Plan 1 Payment Information			
Yearly Income when Starting Payments	Initial Minimum Monthly Payment	Estimate of Total Amount Paid*	Estimate of Length of Time Making Payments*
\$10,000	\$0	\$1,4897.39	20 Years
\$20,000	\$0	\$26,131.91	20 Years
\$30,000	\$83.33	\$51,945.90	18 Years 9 Months
\$40,000	\$166.67	\$42,331.96	12 Years 7 Months
\$50,000	\$250	\$38,334.80	9 Years 3 Months
\$60,000	\$333.33	\$36,233.53	7 Years 3 Months
\$70,000	\$416.67	\$34,959.78	5 Years 11 Months
\$80,000	\$500.00	\$34,112.72	5 Years
\$90,000	\$583.33	\$33,511.35	4 Years 4 Months
*These estimates assume that income increases 5% a year and that the minimum monthly payment is made on the first day of each month			

Repayment Plan 2 Payment Information			
Yearly Income when Starting Payments	Initial Minimum Monthly Payment	Estimate of Total Amount Paid*	Estimate of Length of Time Making Payments*
\$10,000	\$318.20	\$38,184.00	10 Years
\$20,000	\$318.20	\$38,184.00	10 Years
\$30,000	\$318.20	\$38,184.00	10 Years
\$40,000	\$318.20	\$38,184.00	10 Years
\$50,000	\$318.20	\$38,184.00	10 Years
\$60,000	\$318.20	\$38,184.00	10 Years
\$70,000	\$318.20	\$38,184.00	10 Years
\$80,000	\$318.20	\$38,184.00	10 Years
\$90,000	\$318.20	\$38,184.00	10 Years
*These estimates assume that income increases 5% a year and that the minimum monthly payment is made on the first day of each month			

In the survey these tables appear directly below the table in Figure A5.

The \$318.20 monthly payment is calculated using the bankrate.com loan calculator (URL: <https://www.bankrate.com/calculators/mortgages/loan-calculator.aspx>).

The monthly payment is for a \$30,000 loan with a loan term of 10 years at a 5% interest rate. The estimate of Total Amount Paid = \$318.20 payment * 120 monthly payments over 10 years.

The 5% increase in income comes from the U.S. Department of Education’s Loan Simulator. On the page where individuals provide their yearly salary and how much their incomes grow each year the default income growth is 5%. On that page is written “*According to a U.S. Department of Education and U.S. Department of Treasury analysis of a representative sample of actual student loan borrower incomes, the borrower incomes increase, on average, at a rate of 5% per year.” See <https://studentaid.gov/loan-simulator/repayment/wizard/personal-info/income-info> accessed January 20th, 2022.

Figure A7 - Repayment Plan Choice Question



Which repayment plan would you choose?


Repayment Plan 1

Repayment Plan 2

→

In the survey Figure A7 appears directly below Figure A6.

Figure A8 - Test of Understanding of Repayment Plans Introduction


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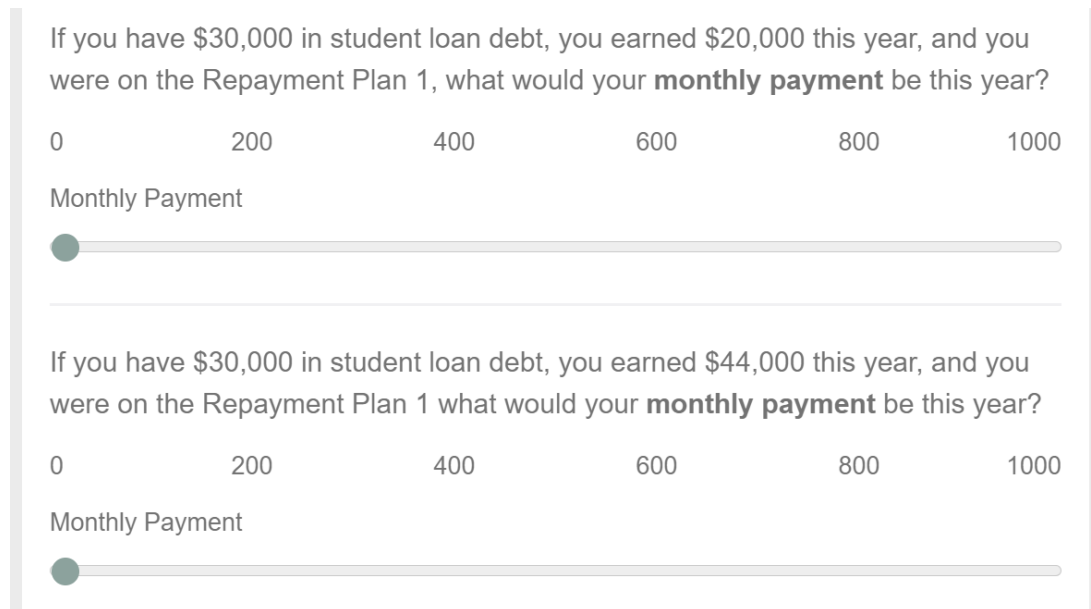
	Minimum Monthly Payment	Loan Forgiveness	Payments Change when Income Changes	Must Pay When Earn Less than \$20,000 a year	Must Pay When in Education	Must Pay When Unemployed
Repayment Plan 1	10% of your annual income above \$20,000 split into 12 monthly payments	Any remaining loan balance after 20 years of payments is forgiven	Yes	No	No	No
Repayment Plan 2	Payments are set so the loan is paid off in 10 years	No loan forgiveness	No	Yes	No	No

Instructions

Please use a **calculator** to answer the following questions.

The slider below will allow you to pick one of 6 possible answers: \$0, \$200, \$400, \$600, \$800, \$1000. The correct answer to each of the next 4 questions is one of those 6 options.

Figure A9 - Test of Understanding of Repayment Plans Questions 1 and 2



In the survey Figure A9 appears directly below Figure A8.

The answer to Question 1 is \$0. The description of the plans in Figure A12 says the borrower does not need to make payments if they are on Repayment Plan 1 and they make less than \$20,000 per year. 41% of respondents answered this question correctly.


The answer to Question 2 is \$200. This is the formula to calculate a monthly payment on Repayment Plan 1 given an annual income above \$20,000: $0.1 * (\text{Annual Income} - \$20,000) / 12$. Plugging \$44,000 for Annual Income into the formula the expression equals \$200. 38% of respondents answered this question correctly.

Figure A10 - Test of Understanding of Repayment Plans Questions 3 and 4

Assume you have student loan debt and you are on **Repayment Plan 2** (different repayment plan from previous question). Your income decreased from \$68,000 last year to \$44,000 this year. Last year your minimum monthly payment was \$400. What is your **minimum monthly payment** this year?

0 200 400 600 800 1000


Monthly Payment



You have been on Repayment Plan 1. It has been 21 years since you started paying off your student loans. At the end of the 20th year, your balance was \$10,000. Your income is the same for both the 20th year and 21st year. During your 20th year, your minimum monthly payment was \$800. What is your **minimum monthly payment** for your 21st year?

0 200 400 600 800 1000

Monthly Payment



In the survey Figure A10 appears directly below Figure A9.

The answer to Question 3 is \$400. The description of the plans in Figure A8 says if you are on Repayment Plan 2 your payments do not change when your income changes. Therefore, the payment stays at \$400. 66% of respondents answered this question correctly

The answer to Question 4 is \$0. The description of the plans in Figure A8 says “Any remaining loan balance after 20 years of payments is forgiven.” Therefore, at the end of the borrower’s 20th year their \$10,000 was forgiven. Therefore, during the borrower’s 21st year they are not required to make any payments. 39% of respondents answered this question correctly.

Survey respondents see Figures A8, A9, and A10 on the same page. That page comes after survey respondents are asked for their repayment plan choice a second time and before questions about covariates like age and gender.

Appendix B: Summary Statistics

Table B1 - Number of Majors

Number	Frequency	Percent of Analysis Sample
1	1208	86.5%
2	180	12.9%
3	5	0.4%
4 or More	3	0.2%

Table B2 - Gender

Gender	Frequency	Percent of Analysis Sample
Female	826	59.2%
Male	548	39.3%
Other	22	1.6%

Table B3 - Race

Race	Frequency	Percent of Analysis Sample
American Indian or Alaska Native	2	0.1%
Asian	118	8.5%
Black or African American	74	5.3%
Native Hawaiian or Pacific Islander	6	0.4%
White	1135	81.3%
Other	61	4.4%

Table B4 - Student Loans

How much student loan debt the respondent had when they answered the survey	Frequency	Percent of Analysis Sample
\$0	567	40.6%
\$1 - \$10,000	217	15.5%
\$10,001 - \$20,000	226	16.2%
\$20,001 - \$30,000	174	12.5%
\$30,001 - \$40,000	78	5.6%
\$40,001 - \$50,000	48	3.4%
Greater than \$50,000	86	6.2%

Table B5 - Pell Grant

Has the respondent every had a Pell Grant?	Frequency	Percent of Analysis Sample
Yes	428	30.7%
No	968	69.3%

Table B6 - First Generation College Student

Is the respondent a first- generation college student?	Frequency	Percent of Analysis Sample
Yes	267	19.1%
No	1129	80.9%

Table B7 - Summary Statistics Continuous Variables

Variable	Mean	Standard Deviation	Min	Max
Age	21.35	1.76	12	46
Probability will Attend Graduate or Professional School in the Next 20 Years	65.41	31.93	0	100

Appendix C: Balance Tests

Table C1 - Multivariate Regression Test

	Major Specific Income Treatment
Single Major	0.0940** (0.0391)
Is Female	0.0196 (0.0280)
Is White	0.0144 (0.0354)
Has Pell Grant	0.0371 (0.0325)
Is First Generation College Student	-0.0055 (0.0372)
Has Student Loans	-0.0553* (0.0284)
Age	0.0001 (0.0082)
Probability Attend Graduate or Professional School	-0.0000 (0.0004)
N	1,396

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Each coefficient in the above table comes from a single regression using the following estimating equation

$$\begin{aligned}
 \text{Major_Specific_Treatment}_i &= \beta_0 + \beta_1 \text{Single_Major}_i + \beta_2 \text{Is_Female}_i + \beta_3 \text{Is_White}_i + \beta_4 \text{Has_Pell_Grant}_i \\
 &+ \beta_5 \text{Is_First_Gen_Col}_i + \beta_6 \text{Has_Student_Loans}_i + \beta_7 \text{Age}_i + \beta_8 \text{Prob_Grad_School}_i + \epsilon_i
 \end{aligned}$$

Major_Specific_Treatment is a binary variable which equals 1 if the respondent saw the Major Specific Income Treatment. Single Major, Is Female, Is White, Has Pell Grant, Is First Generation College Student, and Has Student Loans are binary variables which equal 1 if the respondent has the attribute in the variable name.

Appendix D: Robustness Check

Table D1 - Effect of Treatment on Probability of Earning \$0 to \$60,000

	Percent Chance of Earning \$0 to \$60,000	Percent Chance of Earning \$0 to \$60,000
Major Specific Treatment	5.2699*** (1.8612)	5.4729*** (1.8042)
Covariates	N	Y
N	1,396	1,396
	Percent Chance of Earning \$0 to \$60,000	
Major Specific Treatment	4.1118	
* After Treatment	(2.5362)	
Covariates	N	
N	2,792	

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. This table shows the results of an the subjective probability (scaled to be between 0 and 100) a survey respondent believed they would earn \$0 to \$60,000 5 years after graduating from MSU regressed on binary variables for the survey respondent seeing the major specific income treatment (Major Specific Treatment) or a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors. The covariates are binary variables for the survey respondent: being female, being white, having only 1 major, having a Pell Grant, being a first-generation college student and having student loans. There are also 2 other covariates: the survey respondent's age and the probability the survey respondent believed they would attend graduate or professional school within 20 years of answering the survey.

Appendix E: Heterogeneity by Having Student Loans

Table E1 - Differences Between Survey Respondents with and Without Student Loans

	Has Student Loans
Single Major	0.0440 (0.0371)
Is Female	0.0314 (0.0265)
Is White	0.0002 (0.0324)
Has Pell Grant	0.2340*** (0.0287)
Is First Generation College Student	0.1146*** (0.0325)
Age	0.0130 (0.0088)
Probability Attend Graduate or Professional School	-0.0004 (0.0004)
N	1,396

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. The above table shows the results of a regression of an indicator variable for the survey respondent having student loans on indicator variables for the student having a single major, being female, being white, having a Pell Grant, and being a first-generation college student and continuous variables for the survey respondent's age and their subjective probability of attending graduate or professional school. All standard errors are robust standard errors.

Has Student Loans Only Sample (829 Survey Respondents)

Table E2 - Effect of Treatment on Change in Probability of Earning a Low Income Has Student Loans

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	9.0887***	-0.0424
* After Treatment	(2.2186)	(0.0479)
N	1,658	1,658

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table, each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 829 * 2 = 1,658$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

No Student Loans Sample (568 Survey Respondents)

Table E3 - Effect of Treatment on Change in Probability of Earning a Low Income No Student Loans

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	4.3709*	0.0181
* After Treatment	(2.4929)	(0.0574)
N	1,136	1,136

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table, each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 568 * 2 = 1,136$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believed they would earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

Models with Interaction Terms

Table E4 - Effect of Treatment on Change in Probability of Earning a Low Income by Having Student Loans

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	4.7178	-0.0606
* Have Student Loans	(3.3369)	(0.0747)
*After Treatment		
N	2,792	2,792

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 1,396 * 2 = 2,792$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment), a binary variable for the respondent having student loans (Have Student Loans), and all possible interaction terms using those 3 binary variables. All standard errors are robust standard errors.

Appendix F: Heterogeneity by Income of Major

Survey respondents are categorized as having a low-income major if the income they would have seen if they had seen the Major Specific Income Treatment was below \$37,400. All other survey respondents are classified as having a high-income major. This cutoff is chosen because it was the median income survey respondents would have seen if they had been shown the Major Specific Income Treatment.

Low-Income Majors (694 Survey Respondents)

Table F1 - Effect of Treatment on Change in Probability of Earning a Low-Income Low-Income Majors

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	13.9117***	0.0206
* After Treatment	(2.6683)	(0.0524)
N	1,388	1,388

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table, each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 694 * 2 = 1,388$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believed they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

High Income Majors (702 Survey Respondents)

Table F2 - Effect of Treatment on Change in Probability of Earning a Low-Income High-Income Majors

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	0.0362	-0.0599
* After Treatment	(1.6340)	(0.0516)
N	1,404	1,404

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 702 * 2 = 1,404$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

Models with Interaction Terms

Table F5 - Effect of Treatment on Change in Probability of Earning a Low-Income by Income of Major

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	13.8754***	0.0805
* Have Student Loans	(3.1288)	(0.0735)
*After Treatment		
N	2,792	2,792

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 1,396 * 2 = 2,792$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed: on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment), a binary variable for the respondent having a low income major (Low Income Major), and all possible interaction terms using those 3 binary variables. All standard errors are robust standard errors.

Appendix G: Treatment Effect on Subjective Probability of Earning \$30,000 to \$60,000

Table G1 - Treatment Effect on Subjective Probability of Earning \$30,000 to \$60,000

	Percent Chance of Earning \$30,000 to \$60,000	Percent Chance of Earning \$30,000 to \$60,000
Major Specific Treatment	-2.7190*	-2.5806*
	(1.3939)	(1.3828)
Covariates	N	Y
N	1,396	1,396
	Percent Chance of Earning \$30,000 to \$60,000	
Major Specific Treatment	-2.9829	
* After Treatment	(1.8282)	
Covariates	N	
N	2,792	

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$30,000 to \$60,000 5 years after graduating from MSU regressed on binary variables for the survey respondent seeing the major specific income treatment (Major Specific Treatment) or a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors. The covariates are binary variables for the survey respondent: being female, being white, having only 1 major, having a Pell Grant, being a first-generation college student and having student loans. There are also 2 other covariates: the survey respondent's age and the probability the survey respondent believed they would attend graduate or professional school within 20 years of answering the survey.

Appendix H: Heterogeneity by Repayment Plan Understanding

Survey respondents are classified as having low plan understanding if they got 0 or 1 questions right on the 4-question test of repayment plan understanding in the survey.

Survey respondents are classified as having high plan understanding if they got 2, 3, or 4 questions right on the 4-question test of repayment plan understanding in the survey.

This way of categorizing survey respondents is chosen to split the sample as evenly as possible.

Low Plan Understanding (672 Survey Respondents)

Table H1 - Effect of Treatment on Change in Probability of Earning a Low-Income Low Plan Understanding

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	7.7501***	0.0117
* After Treatment	(2.5706)	(0.0543)
N	1,344	1,344

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 672 * 2 = 1,344$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

High Plan Understanding (724 Survey Respondents)

Table H2 - Effect of Treatment on Change in Probability of Earning a Low-Income High Plan Understanding

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	6.6260***	-0.0462
* After Treatment	(2.1355)	(0.0489)
N	1,448	1,448

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 724 * 2 = 1,448$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believed they would earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment) and an interaction between those variables. All standard errors are robust standard errors.

Models with Interaction Terms

Table H3 - Effect of Treatment on Change in Probability of Earning a Low Income by Plan Understanding

	Percent Chance of Earning a Low Income	Choose IDR Plan
Major Specific Treatment	1.1241	0.0580
* Have Student Loans	(3.3419)	(0.0731)
*After Treatment		
N	2,792	2,792

Notes: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. For this table each survey respondent has two observations: one observation before the treatment and one after the treatment. Therefore, $N = 1,396 * 2 = 2,792$. This table shows the results of the subjective probability (scaled to be between 0 and 100) a survey respondent believes they will earn \$0 to \$30,000 5 years after graduating from MSU or a binary variable for the survey respondent choosing the IDR plan regressed: on a binary variable for seeing the Major Specific Income Treatment, a binary variable for the question about income expectations coming after the treatment (After Treatment), a binary variable for the respondent having a low amount of understand of the repayment plans (Low Plan Understanding), and all possible interaction terms using those 3 binary variables. All standard errors are robust standard errors.