

Risk Preferences, GPA, and Demographic Factors as Predictors in Student Course Selection

Rachel Behrmann,¹ Dennis Kramer II²

¹Department of Economics, University of Florida, Gainesville, FL
²College of Education, University of Florida, Gainesville, FL

Introduction

- The value of a college education is not simply the knowledge gained, but often includes the resulting employment outcomes for that individual. It is therefore important that students make course-selecting decisions that maximize their skillsets necessary for being successful in the workforce and the job market.
- There is very little research on why students select certain courses and it is a highly understudied topic (Babad, 2001; Babad and Tayeb, 2003).
- Studies have shown that, in addition to course quality, other factors come into consideration in the course selection process as well (Babad, 2001). One consideration in particular, ease, impacts course-selecting behavior, even though it is not directly related to the quality of the course. Another consideration, then, could be that students select courses according to their risk preferences, and the ease of a course should be relevant in this matter, especially when considering its effect on student GPA.
- This study aims to examine how course selection varies with risk preferences, GPA, and other demographic factors. We hypothesize that risk-averse students will be more likely to take easier courses, and that we will see a relationship between students' preferences for ease and their preferences for quality.

Methodology

A Qualtrics survey was administered to an introductory macroeconomics course for extra credit over a two-week period. The survey consisted of three parts:

- Course selection procedure—similar to a study done by Babad and Tayeb (2003), but it modeled course selection options off of information provided by Rate My Professors on ease and quality of the course.
- 5-trial adjusting probability discounting task—measured risk preferences in individuals to determine how risk-seeking/averse they were.
- Demographics—collected gender, year in school, age, plans after school (e.g. industry), and GPA.

We received 652 responses during the specified period. After removing participants who didn't complete the course selection procedure, didn't follow the directions, or selected that they were not paying attention and did not wish to have their data included, 567 responses remained. Of the remaining participants, approximately 58% were female and the remaining 42% were male.

Imagine you are selecting an elective from your major. Please select your top three choices out of the nine below. Level of difficulty is ranked from 1-least difficult to 5-most difficult. Overall quality is ranked from 1-least quality to 5-highest quality.

Option	Section	Level of Difficulty	Overall Quality
1	1256	4.5	4.5
2	4488	4.5	2.5
3	2784	4.5	1
4	3172	2.5	4.5
5	2467	2.5	2.5
6	3291	2.5	1
7	2785	1	4.5
8	3542	1	2.5
9	1724	1	1

Please select your top three courses based on level of difficulty and overall quality. Make sure to select three different courses.

1st Choice: \$50 for sure
 2nd Choice: \$100 with a 50% chance
 3rd Choice:

What would you prefer?
 \$50 for sure
 \$100 with a 50% chance

This is the course selection procedure students completed.

Probability discounting task. After the first question, the probability adjusted up or down according to the student's response.

Index	Uncertain Choice	Olds Against	Trial No.	Certain	Uncertain	Certain	Uncertain
1	0.97	0.0309	5	0.0309	0.0443	32.3333	22.50679
2	0.94	0.0618	4	0.0618	0.11613	12.5860	8.61093
3	0.91	0.0927	5	0.0927	0.17133	12.5860	8.61093
4	0.88	0.1236	3	0.1236	0.22653	12.5860	8.61093
5	0.85	0.1545	5	0.1545	0.28173	12.5860	8.61093
6	0.82	0.1854	4	0.1854	0.33693	12.5860	8.61093
7	0.79	0.2163	5	0.2163	0.39213	12.5860	8.61093
8	0.76	0.2472	3	0.2472	0.44733	12.5860	8.61093
9	0.73	0.2781	5	0.2781	0.50253	12.5860	8.61093
10	0.70	0.3090	4	0.3090	0.55773	12.5860	8.61093
11	0.67	0.3399	5	0.3399	0.61293	12.5860	8.61093
12	0.64	0.3708	3	0.3708	0.66813	12.5860	8.61093
13	0.61	0.4017	5	0.4017	0.72333	12.5860	8.61093
14	0.58	0.4326	4	0.4326	0.77853	12.5860	8.61093
15	0.55	0.4635	5	0.4635	0.83373	12.5860	8.61093
16	0.52	0.4944	3	0.4944	0.88893	12.5860	8.61093
17	0.49	0.5253	5	0.5253	0.94413	12.5860	8.61093
18	0.46	0.5562	4	0.5562	0.99933	12.5860	8.61093
19	0.43	0.5871	5	0.5871	1.05453	12.5860	8.61093
20	0.40	0.6180	3	0.6180	1.10973	12.5860	8.61093
21	0.37	0.6489	5	0.6489	1.16493	12.5860	8.61093
22	0.34	0.6798	4	0.6798	1.22013	12.5860	8.61093
23	0.31	0.7107	5	0.7107	1.27533	12.5860	8.61093
24	0.28	0.7416	3	0.7416	1.33053	12.5860	8.61093
25	0.25	0.7725	5	0.7725	1.38573	12.5860	8.61093
26	0.22	0.8034	4	0.8034	1.44093	12.5860	8.61093
27	0.19	0.8343	5	0.8343	1.49613	12.5860	8.61093
28	0.16	0.8652	3	0.8652	1.55133	12.5860	8.61093
29	0.13	0.8961	5	0.8961	1.60653	12.5860	8.61093
30	0.10	0.9270	4	0.9270	1.66173	12.5860	8.61093
31	0.07	0.9579	5	0.9579	1.71693	12.5860	8.61093
32	0.04	0.9888	3	0.9888	1.77213	12.5860	8.61093

As a measure of risk preferences, h values were assigned based on the last choice in the probability discounting task (Cox and Dallery, 2016).

Results

In analyzing this data, we ran three regressions to model 1) variables in ease selections, 2) variables in quality selections, and 3) the relationship between selecting for ease and quality. As a robustness check, it was suggested that we run a separate regression on ease and quality selections to account for unusual choices.

	SS	df	MS
Model	3.033	9	0.337
Residual	146.750	558	0.263
Total	149.783	567	0.266

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
logh	0.015	0.047	0.33	0.741	-0.076 0.107
male = 1	0.088	0.044	2.00	0.046	0.002 0.174
year	-0.056	0.024	-2.36	0.019	-0.102 -0.009
industry = 2	-0.032	0.050	-0.64	0.520	-0.131 0.066
other = 3	0.076	0.068	1.12	0.263	-0.058 0.210
cons	1.891	0.056	34.03	0.000	1.782 2.000

	SS	df	MS
Model	0.493	5	0.099
Residual	68.310	558	0.122
Total	68.803	563	0.122

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
qualityrating	0.015	0.047	0.33	0.741	-0.076 0.107
male = 1	-0.001	0.030	-0.05	0.962	-0.060 0.058
year	-0.002	0.016	-0.15	0.883	-0.034 0.029
industry = 2	0.006	0.034	0.16	0.869	-0.062 0.073
other = 3	-0.059	0.046	-1.27	0.205	-0.150 0.032
cons	2.832	0.038	74.69	0.000	2.758 2.907

In these final models, ease and quality preferences were regressed against their h scores (risk preferences), year in school, gender, and future plans (graduate school, industry, or other). Only gender and year (specifically for ease preferences) proved significant in this analysis.

	SS	df	MS
Model	9.190	7	1.313
Residual	141.128	558	0.253
Total	150.317	565	0.266

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
logh	0.361	0.148	2.45	0.015	0.071 0.651
male = 1	0.368	0.158	2.33	0.020	0.058 0.679
year	-0.171	0.081	-2.11	0.035	-0.331 -0.012
not "unusual" = 1	-0.262	0.194	-1.35	0.177	-0.642 0.118
dummy*logh	-0.389	0.155	-2.51	0.012	-0.694 -0.084
dummy*gender	-0.312	0.165	-1.90	0.058	-0.635 0.011
dummy*year	0.126	0.085	1.49	0.138	-0.040 0.292
cons	2.140	0.185	11.54	0.000	1.776 2.504

	SS	df	MS
Model	47.119	7	6.731
Residual	21.756	558	0.039
Total	68.875	565	0.122

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
logh	-0.140	0.058	-2.50	0.010	-0.268 -0.016
male = 1	-0.002	0.062	-0.04	0.971	-0.124 0.120
year	0.022	0.032	0.68	0.495	-0.041 0.084
not "unusual" = 1	1.099	0.076	14.47	0.000	0.950 1.249
dummy*logh	0.171	0.061	2.80	0.004	0.050 0.291
dummy*gender	-0.002	0.065	-0.03	0.977	-0.129 0.125
dummy*year	-0.028	0.033	-0.85	0.396	-0.093 0.037
cons	1.810	0.073	24.87	0.000	1.667 1.953

A dummy variable was interacted with the other variables to see what effect "unusual" choices had on the data. Here, an unusual choice was defined to be selecting anything other than the highest quality option for the first course (7.41% of the sample).

Variable	Obs	Mean	St. Dev.	Min	Max
easeringating	567	1.830	0.516	1	3
qualityrating	567	2.803	0.349	1	3
logh	567	0.390	0.466	-1.38	1.51
age	559	19.390	2.613	16	47
year	566	1.859	0.928	1	5
gpa	323	3.413	0.431	1.8	4

Future Plans	Freq.	Percent	Cum.
Graduate School	340	60.18	60.18
Industry	155	27.43	87.61
Other	70	12.39	100
Total	565	100	

Sample Demographics.

Option 1	Option 2	Option 3	Option 4	Option 5	Option 6	Option 7	Option 8	Option 9
15.34%	1.94%	1.94%	36.86%	0.71%	0.71%	40.39%	0.88%	1.23%

Distribution of first choice course selections.

	SS	df	MS
Model	3.943	1	3.943
Residual	146.595	565	0.259
Total	150.538	566	0.266

	Number of obs
F(1, 565)	15.2
Prob > F	0.000
R-squared	0.026
Adj R-squared	0.025
Root MSE	0.509

Regression of ease preferences with quality. Here, preferences for higher quality correlate with easier courses.

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
easeringating	-0.239	0.061	-3.9	0.000	-0.360 -0.119
qualityrating	2.501	0.173	14.42	0.000	2.161 2.842
cons					

Conclusions

- The evidence does not support the hypothesis that student course selection is correlated with risk preferences.
- In terms of selection for ease, males were more likely to choose harder courses and students tended to increase in their preferences for ease as they increased in year in school.

Future Directions

Future studies should attempt to:

- Replicate these results and
- Improve modeling the course selection process in order to better understand and study the factors that contribute to course selecting decisions

References

- About RateMyProfessors.com. (n.d.). Rate My Professors. Retrieved from <http://www.ratemyprofessors.com/About.jsp>
- Babad, E. (2001). Students' course selection: Differential considerations for first and last course. *Research in Higher Education*, 42(4), 469-492. Retrieved from JSTOR Journals
- Babad, E., & Tayeb, A. (2003). Experimental analysis of students' course selection. *British Journal of Educational Psychology*, 73(3), 373-393. doi:10.1348/000709903322275894
- Cox, D. J., & Dallery, J. (2016). Effects of delay and probability combinations on discounting in humans. *Behavioral Processes*, 131, 15-23. doi:10.1016/j.beproc.2016.08.002

Acknowledgements

The authors would like to thank David Cox, Perihan Saygin, PhD, Molly Barlow, and Maria Ripol for their advice, guidance, and assistance during this process.