

Investigating Student Motivation and Performance in Electrical Engineering and Its Subdisciplines

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Abstract—Factors influencing choice of major in electrical engineering and later curricular and professional choices are investigated. Studies include both quantitative and qualitative analyses via student transcripts, surveys, and focus groups. Student motivation for choosing an electrical engineering major and later subdiscipline in the field is interpreted through expectancy-value theory, where primary factors of strong perceived value of future professional opportunities and strong influence of course instructors are identified. Performances in select required electrical engineering courses appear to serve as predictors for student choice of subdiscipline emphasis. In contrast, participation in student professional activities does not show statistically significant correlations with subdiscipline. Curricular and professional choices appear to be explained by expectancy-value theory with inclusion of socializers. The findings suggest that early and integrative exposure of all electrical engineering technical areas, including high-quality teaching, may provide an optimal basis for students to make future decisions on academic path and participation in professional activities.

Index Terms—Choice of major, expectancy-value theory, social influences, subdiscipline.

I. INTRODUCTION

ELECTRICAL engineering (EE) is one of the broadest disciplines within engineering, with its subfields of study often being unknown to students in their early undergraduate years. While many students are familiar with electrical power, circuits, and electromagnetics, key areas including signal processing, control systems, optics, and semiconductor devices are often unfamiliar. In addition to academic performance, the student experience extends well beyond the classroom, where engaging in the multitude of research projects, student project teams, and related professional experiences can transform educational outcomes. Identifying factors that correlate with students' motivation to choose EE as a major, their choice of subdiscipline within EE, and their academic performance is invaluable in informing recruiting, advising, mentoring, and curricular reform activities.

Prior studies have examined factors that correlate to choice of major within engineering [1]–[5]. A study drawing from a large multi-institution database concluded that gender and race are important factors [1]. For example, the study found that EE is much more popular with Asian and Black females than Hispanic and White females, recommending further study to understand the reasons behind the enrollment and graduation statistics. Differences in gender and parental educational

achievement were cited as contributing factors for choice of engineering major at the University of New Haven, West Haven, CT, USA, recommending further inclusion of qualitative data and larger sample size [2]. Another analysis of a national multi-institutional survey indicated that the level of motivation for students of different engineering majors may vary with gender, but that the variation is not significant for EE [3]. Factors determining choice of an engineering major have been studied previously [4], where a qualitative study based on expectancy-value theory emphasized the importance of understanding values and connecting personal identities to engineering identities. The diversity of technical fields within engineering, and subdisciplines within a particular engineering major, however, may attract student populations with differing values and identities.

The diversity of technical areas within EE may incite varied motivations, attitudes, and expectations among students; gaining an understanding of these facets may be used to guide educational reforms. Studies of motivations and attitudes in various engineering majors have been shown to exhibit different expectancies of success for student performance in their major; understanding these differences is believed to help direct instructional change to suit the student population within a given major [5]. Instructional change has also been shown to impact recruiting of mechanical engineering students to pursue a minor in EE at Temple University, Philadelphia, PA, USA [6], where emphasis on pedagogical strategy and motivation were shown to be effective. EE can be considered, and taught, as a set of subdisciplines, but it may ultimately be better to develop a more holistic and interconnected teaching of EE material between subdisciplines. Initiatives at Duke University, Durham, NC, USA, have reported advantages in following this approach [7]. Understanding student motivation and success at the University of Michigan, Ann Arbor, MI, USA, is an important step towards undertaking instructional reform that can better address student needs.

Student motivations to major in EE, and the choices they make during their education, can be interpreted through the lens of expectancy-value theory, including the impact of social influences [8]–[11]. The expectancy-value theory describes decision making as based on the individual's expectancy of success and on the anticipated value of the outcome, which includes negative value (costs) of engaging in the action. With respect to choosing EE as a major, students' perceptions of their ability to graduate in the field, and the value they see in career opportunities in EE, will affect their decisions. These motivations include the influence of social factors on achievement-related motivations and choices [9].

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97 Ultimately, to support success for students in EE, an un-
 98 derstanding is needed of why they enroll in the first place
 99 and how they progress through the program. Thus, this work
 100 focuses on students' motivations to become an EE, and on how
 101 student characteristics and choices correlated to success in an
 102 EE program. This project was undertaken in the Department
 103 of Electrical Engineering and Computer Science (EECS) at the
 104 University of Michigan to understand factors that determine
 105 student choices for major, subdiscipline, and participation in
 106 professional activities. The information garnered from the study
 107 could then be used to target recruiting activities, improve stu-
 108 dent advising, and motivate future initiatives in the department
 109 for curricular reform and extracurricular professional activities.
 110 This project examines academic performance and behavior for
 111 students in EE, as well as subdiscipline trends within EE.
 112 Data are analyzed to study potential dependencies on student
 113 performance, choice of classes, involvement in professional
 114 activities outside of the classroom, and overall impressions
 115 of the undergraduate experience. Historical academic perfor-
 116 mance data, feedback from focus groups, and survey data from
 117 graduating students over the past two years are used to examine
 118 motivations for choosing EE and curricular and co-curricular
 119 choices in EE.

120

II. METHODS

121 This study is guided by two research questions : 1) What
 122 reasons do students cite for choosing EE as a major?; and
 123 2) How does student performance in EE vary by student
 124 demographics and curricular and co-curricular choices? Data
 125 from student transcripts, surveys of graduating students, and
 126 focus groups of current students are used to address these ques-
 127 tions. To provide some context to the population of this study,
 128 undergraduate students enroll in the College of Engineering,
 129 and later declare their major within engineering, typically in
 130 freshman or sophomore year. The College of Engineering is
 131 highly selective; students entering in the Fall 2014 semester had
 132 a 28% acceptance rate, mean high school grade point average
 133 of 3.9, mean ACT score of 33, and mean SAT score of 1430.

134 Transcript data were analyzed for students graduating with
 135 a bachelor's degree in EE between 2005 and 2012. The data
 136 included grade performance, gender, and choice of academic
 137 courses, and provided information on student performance, de-
 138 mographics, and curricular choices. The data set encompasses
 139 1032 graduates and includes grades for core math, physics, and
 140 engineering courses required in EE, as well as the selected
 141 senior capstone design courses. Descriptions of the courses
 142 are shown in Table I. In the EE program at the University of
 143 Michigan, students choose their senior capstone design courses
 144 from a set of courses with varying technical emphasis. At least
 145 one of the chosen course combinations includes an EE prereq-
 146 uisite elective course. The choice of capstone design course
 147 provides an indicator for subdiscipline (though there are no
 148 formal declaration requirements). For the purpose of this study,
 149 subdisciplines are categorized in the areas of circuits/solid state,
 150 electromagnetics/optics, signals and systems, and computers.

151 Surveys were constructed to be administered to graduating
 152 students to investigate student behavior not available on student

TABLE I
 LIST OF COURSES EXAMINED IN THE 2005–2012 DATA SET
 USED IN THIS STUDY

Course	Title	Subdiscipline
Math 215	Calculus III	Required
Math 216	Differential Equations	Required
Physics 240	General Physics II	Required
Engineering 100	Intro. Engineering	Required
EECS 215	Electronic Circuits	Required
IEECS 216	Signals and Systems	Required
EECS 230	Electromagnetics I	Required
EECS 280	Programming	Required
EECS 320	Semiconductor Devices	Required
EECS 401	Probability	Required
EECS 411	Microwave Circuits	Circuits/Solid State
EECS 413	Amplifier Circuits	Circuits/Solid State
EECS 425	Integrated Microsystems	Circuits/Solid State
EECS 427	VLSI Design	Circuits/Solid State
EECS 430	Radiowave Propagation	EM/Optics
EECS 438	Lasers and Optics	EM/Optics
EECS 452	Digital Signal Processing	Signals/Systems
EECS 470	Computer Architecture	Computers

transcripts (e.g., participation in professional activities), and to
 153 inquire directly about students' motivations for majoring in EE.
 154 The survey instrument, based on prior studies of engineering
 155 major choice [2] and choice of major in Information Systems
 156 [12], was adapted for the academic program at the University
 157 of Michigan and posted at [13]. Likert-scale questions used
 158 a six-point scale to eliminate the ability to select a neutral
 159 response. The survey was offered to students through the EECS
 160 undergraduate advising office when they submit paperwork for
 161 graduation. The results shown in this study are for students
 162 graduating in the 2013–2014 and 2014–2015 academic years
 163 (a subset of the pool of students covered by the transcript data
 164 described earlier). A total of 163 students (62% of students
 165 receiving a bachelor's degree in EE during this timeframe)
 166 responded. 167

168 Additionally, focus groups were conducted to provide addi-
 169 tional qualitative information on how students decide to choose
 170 EE as a major and on other choices they make in their academic
 171 career. Four focus groups were conducted: for students who
 172 had not yet declared a major, for recently declared majors,
 173 for juniors, and for seniors who had completed their major
 174 design course. Ten randomly chosen students were contacted
 175 by e-mail from targeted courses to provide representative data
 176 for the intended student categories: ENGR 110 (survey course
 177 in engineering for first year students with undecided major),
 178 EECS 215, EECS 320, and EECS 496 (required course on engi-
 179 neering professionalism to accompany senior capstone design).
 180 The aim was to capture a cross section of students at their
 181 first, second, third, and fourth year, regardless of subdiscipline.
 182 Students could decline the invitation. Each focus group had
 183 between five and eight participants, was conducted by a single
 184 researcher, and lasted about an hour. Focus group discussion
 185 was audio recorded for further analysis. Students were asked 185

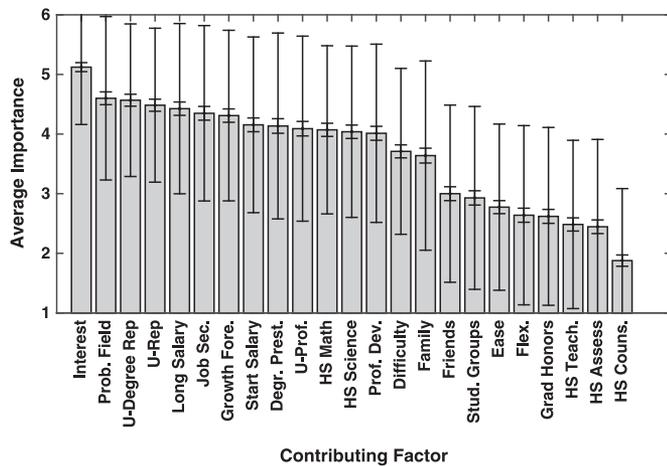


Fig. 1. Ranked average importance for source of influence in decision to choose major in electrical engineering. Values are shown for the mean response in each category, standard deviation (large error bars), and standard deviation on the mean (small error bars). (Prob. Field = probability working in the field after graduation.)

186 approximately five questions related to their experiences and
 187 choices. Sample questions included “How did you choose
 188 EE?”, “Who or what influenced your decision?”, and “How did
 189 you choose your senior capstone design course?”.

190

III. FINDINGS

A. Motivations

192 The decision to choose EE as a major was investigated using
 193 data from surveys and focus groups. The average importance
 194 of contributing factors to choice of major in EE is shown in
 195 Fig. 1 in ranked order. Interest in the technical area of EE, job
 196 prospects and related salary, and the prestige of the institution
 197 and degree were rated highly as contributing factors to the
 198 choice of EE as a major. Influences at the high school level,
 199 the perceived easiness of technical field and job flexibility, and
 200 influence of family and friends were rated with less emphasis.
 201 The survey data suggest that students who eventually choose
 202 EE, and are ultimately on the path to graduation, value their
 203 technical interests in the field and their career prospects higher
 204 than social influences. These motivations for choosing EE as
 205 a discipline agree well with expectancy-value theory, where
 206 career prospects and inherent technical interest provide the
 207 largest combination of expected success and enjoyment (value)
 208 for these students. Perhaps not surprisingly, graduating students
 209 also reported that they were well informed about the technical
 210 field and career prospects in EE prior to declaring, and even
 211 more aware of the field of EE than other engineering fields.
 212 A comparative survey for students who choose engineering
 213 majors in other fields would provide a better understanding of
 214 the importance of prior knowledge of a particular engineering
 215 discipline as a contributing factor to choice of major.

216 Focus group responses indicated that students were often
 217 inherently interested in the subjects of EE, with their parents
 218 playing a role in their understanding of the versatility of the
 219 major. For example, one respondent reported, “My dad did his
 220 undergrad in EE. He said if you don’t know what to do, try EE.

221 *The analytical analysis is good in general.*” Parental influence
 222 and the importance of general analytical skills in choice of EE
 223 major were reported by multiple respondents, which is also
 224 consistent with general feedback from EE advisors. This data
 225 suggests the impact of socializers on choice of engineering
 226 major, consistent with expectancy-value theory with parents
 227 acting as a model for their children.

228 To understand how EE students then choose a specialization
 229 within the discipline, feedback was analyzed from focus group
 230 discussions. In some instances, students indicated that their
 231 experiences in research or at an internship motivated them to
 232 enroll in classes associated with their work. Even with nontech-
 233 nical internships, the introduction to a specific subdiscipline
 234 motivated student interest in the field. For example, one stu-
 235 dent describing experience at a summer internship commented,
 236 *“I took power systems because I saw the power team at*
 237 *Northrop Grumman was interested (in the topics covered by the*
 238 *course). I thought if I took a power class, I might be able to*
 239 *work with the power team.”* Furthermore, many students indi-
 240 cated that particular professors or graduate student instructors
 241 played significant roles in motivating students’ appreciation of a
 242 subdiscipline, as evidenced by students subsequently enrolling
 243 in classes taught by, or suggested by, the instructors. One early
 244 required course, in particular, is taught by two instructors who
 245 have different teaching styles. One of these professors has
 246 motivated students to pursue the course’s subdiscipline, while
 247 the other has caused students to not only move away from
 248 that subdiscipline, but to even change their major to computer
 249 science. Another student describing their choice of EECS 411
 250 as their senior capstone design course reported, *“I wanted to*
 251 *take that after taking 230 with Prof. X. I liked the transmission*
 252 *line stuff. The courses that led up to that were 230, 311 analog*
 253 *circuits (Prof. Y was very good). I really understood those*
 254 *classes well.”* While this is data garnered from a small number
 255 of students, it illustrates the impact instructor reputation and
 256 ability may have on student motivation. Open-form responses
 257 on the surveys also suggest that specific instructors were highly
 258 influential on student decisions on courses and activities.

259 The impact individuals can have on student choice of a
 260 subdiscipline is illustrated several times by this data, where
 261 students are profoundly affected by social interaction with
 262 experts in the field. This information suggests that having
 263 successful instructors teach introductory courses may lead to
 264 higher student motivation and better retention within EE.

265 The motivations for choosing EE as a major as well as
 266 subdiscipline within EE appear to follow expectancy-value
 267 theory, with the initial choice of major seeming to be largely
 268 dictated by the prospect of future professional opportunities and
 269 inherent interest in the field. Students also report that outside
 270 influences, including parents and particular course instructors,
 271 can strongly influence decisions, suggesting the influence of
 272 socializers on expectancy-value and motivating choice of major
 273 and subdiscipline in EE.

B. Performance

274 Student performance was also studied to gain a better un-
 275 derstanding of how students succeed in EE. The relationship 276

TABLE II
ANOVA ANALYSIS TO TEST ASSOCIATION OF GPA WITH LISTED
DEPENDENT VARIABLES

Dependent variable	<i>p</i> -value
Gender	.486
Parental Educational Background	.114
Plan to attend grad school in science/engineering	<.001
Plan to begin job in engineering	<.001
Involvement in student society	.138
Involvement in student project team	.689
Involvement in research	.005
Involvement in internship/co-op	.061

TABLE III
ANOVA ANALYSIS TO TEST ASSOCIATION OF RELATIVE PERFORMANCE
IN A PARTICULAR COURSE WITH SUBDISCIPLINE
IN ELECTRICAL ENGINEERING

Course	<i>p</i> -value
Math 215 - Calculus III	.740
Math 216 - Differential Equations	.114
Physics 240 - General Physics II	.675
EECS 215 - Circuits	.001
EECS 216 - Signals and Systems	.002
EECS 230 - Electromagnetics I	<.001
EECS 280 - Programming	.001
EECS 320 - Semiconductor Devices	<.001
EECS 401 - Probability	.045

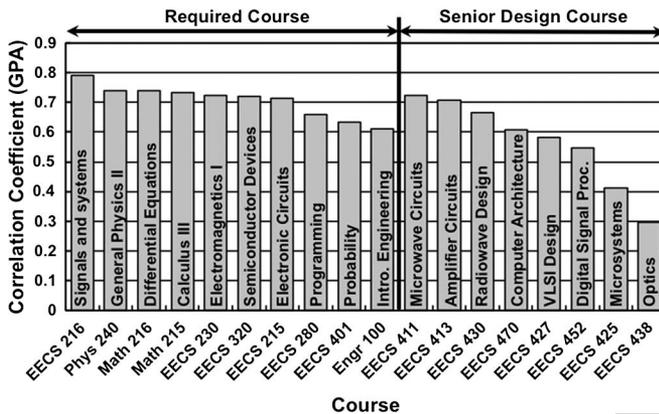


Fig. 2. Correlation coefficient between graduating GPA and grade in selected required courses and senior capstone design course.

277 between student performance and demographics was exam-
278 ined from the survey data containing mean self-reported grade
279 point average (GPA), gender, parental background, and plans
280 after graduation. Differences between group performance were
281 examined using a *t*-test statistical analysis, where *p*-values
282 less than 0.05 suggest statistical significance. The comparative
283 groups and associated *p*-values are summarized in Table II. The
284 *p*-value results suggest that there is not a probable correlation of
285 GPA with gender or parental educational background (whether
286 or not a parent has a college degree). However, there is a
287 highly probable correlation of GPA with future plans to attend
288 graduate school and to work in a job in engineering directly
289 following graduation. The survey data suggest that students
290 with high GPA intend to pursue graduate studies in science and
291 engineering, while lower-achieving students plan to begin a job
292 in engineering directly following graduation.

293 Potential predictors of academic success in EE were stud-
294 ied using student transcript data to determine the degree of
295 correlation between graduating GPA and course grade in early
296 required courses. Correlation coefficients were examined for all
297 required courses in EECS, and a selection of required courses
298 in math, physics, and engineering. A summary of calculated
299 correlation coefficients for courses with respect to graduating
300 GPA is shown in Fig. 2, sorted in descending order and by
301 category of required course or senior capstone design. Strong
302 correlations are observed for required courses, while weaker
303 correlations are observed for the major design capstone courses,

likely due to relatively high grades for all students in these team 304
design courses. As EECS 215 is the first required course for EE 305
within the major, it is not surprising to see a strong correlation 306
and the probability that the course serves as a predictor of 307
future academic success in the major. In contrast, the EECS 308
438 course is a senior capstone design course where most 309
students and their groups succeed in their open-ended design 310
challenge, consequently receiving relatively high grades for all 311
students. The EECS 280 course demonstrates a relatively strong 312
correlation, though the average grade is shifted below the one- 313
to-one line, indicating generally lower relative performance or 314
a “grade penalty” in this course on computer programming. 315
The lower relative performance may be indicative of weaker 316
performance of EE students in computer programming relative 317
to Computer Science majors (it should be noted that the average 318
grade for EECS 280 is consistent with 200-level EECS courses 319
including EECS 215 and EECS 216). 320

Correlations between choice of subdiscipline in EE (defined 321
by chosen senior capstone course) on performance in early 322
required courses were also studied. The relative performance 323
in the required courses (course grade relative to graduating 324
GPA, on a 4-point scale) were examined for groups of students 325
in different subdisciplines as identified by choice of senior 326
capstone design according to 327

$$\Delta_{\text{GPA}} = \text{GPA}_{\text{class}} - \text{GPA}_{\text{graduation}} \quad (1)$$

where Δ_{GPA} is the relative performance metric of a particular 328
subdiscipline, $\text{GPA}_{\text{class}}$ is the grade associated with that class 329
for that subdiscipline, and $\text{GPA}_{\text{graduation}}$ is the graduating 330
GPA. The Δ_{GPA} provides a way of comparing class perfor- 331
mance across subdisciplines to a score that is normalized to 332
graduating GPA. The statistical significance of relative perfor- 333
mance in required courses was compared for varying groups 334
using ANOVA analysis, as shown in Table III. These groups 335
included the subdisciplines defined above. No statistically sig- 336
nificant differences between groups for the required math and 337
physics courses were observed. However, all of the required 338
EECS courses demonstrated a statistically significant differ- 339
ence between groups, indicated with *p*-values < 0.05. The low 340
probability for association of relative performance in math and 341
physics courses with subdiscipline suggests that these courses 342

TABLE IV
SCHEFFE POST-HOC ANALYSIS TO COMPARE RELATIVE PERFORMANCE
IN REQUIRED EECS COURSES FOR STUDENTS IN ONE GROUP TO
STUDENTS IN OTHER GROUPS

Other Design		
EECS 215	Δ_{GPA}	<i>p</i> -value
Circuits/Solid-State	-0.253	.004
Electromagnetics/Optics	-0.160	.271
Signals/Systems	-0.232	.012
Computers	-0.244	.027
EECS 230		
Circuits/Solid-State	-0.234	.009
Electromagnetics/Optics	-0.319	<.001
Signals/Systems	-0.198	.044
Computers	-0.244	.019
Computers		
EECS 280	Δ_{GPA}	<i>p</i> -value
Circuits/Solid-State	0.233	.009
Electromagnetics/Optics	0.291	<.002
Signals/Systems	0.183	.056
Other Design	0.252	.143
Circuits/Solid State		
EECS 320	Δ_{GPA}	<i>p</i> -value
Electromagnetics/Optics	0.150	.012
Signals/Systems	0.128	.004
Computers	0.164	.009
Other Design	0.188	.049
Signals/Systems		
EECS 216	Δ_{GPA}	<i>p</i> -value
Circuits/Solid-State	0.067	.840
Electromagnetics/Optics	0.250	.007
Computers	0.080	.910
Other Design	0.274	.213
EECS 401		
Circuits/Solid-State	0.133	.191
Electromagnetics/Optics	0.086	.759
Computers	-0.007	1.000
Other Design	0.203	.383

do not have a clear impact on later curricular choices. However, all of the required EECS courses demonstrate a statistically significant association of relative course performance with subdiscipline, highlighting the potential impact of these courses on later curricular choices.

Multiple comparisons were examined using Scheffe post-hoc analysis for all required courses that demonstrated statistically significant differences ($p < 0.05$) between groups (all required EECS courses), as shown in Table IV. The analysis compares student performance in courses between groups defined according to choice of senior capstone design (subdiscipline). The analysis highlights cases where significance is probable, warranting further attention and discussion of behavior or per-

TABLE V
CHI-SQUARE ANALYSIS TO TEST ASSOCIATION OF GENDER OR
SUBDISCIPLINE LISTED DEPENDENT VARIABLES

Participatory activity for dependency test	Gender <i>p</i> -value	Subdiscipline <i>p</i> -value
Student organization	.693	.690
Student project team	.693	.557
Research project	.069	.491
Internship/co-op	.405	.003

formance for particular subdisciplines in EE. These cases are described in the following sections.

Senior Capstone Design Outside of EE: Students may choose a senior capstone design course outside of EECS, which may be a multidisciplinary design project or senior capstone design from another department. These students exhibit relative performance that is statistically lower than other students in EECS 215 and EECS 230; these students perform approximately 0.25 grade points below other groups in these two classes. The trend in student performance when choosing a capstone design outside of EE may suggest students decide to pursue other directions because of lower performance compared to their peers.

Students in Computers, Circuits/Solid State: Students choosing a capstone design course in computers demonstrate higher relative performance in the EECS 280 programming course, while students choosing a capstone design course in circuits/solid state demonstrate higher relative performance in the EECS 320 semiconductor device course, relative to other students in EE. These indicate that success and relative performance in EECS 280 and EECS 320 may be strong predictors for choice of subdiscipline in EE.

Signals and Systems: The EECS 216 signals and systems course and EECS 401 course on probabilistic methods represent the required courses in the signals and systems area. While increased relative performance was observed in these courses for students choosing a capstone design course in signals and systems, the ANOVA analysis did not suggest statistical significance. One possible explanation is that the signals and systems capstone design course is considered the most general or least specialized of the EECS senior design courses to choose from, and may be disconnected from interest or performance in a given subdiscipline.

Comparisons of the subdisciplines show strong correlations for select subdiscipline and course combinations such as circuits/solid state and EECS 320. For these combinations, course performance may be highly influential in student motivation for choosing an EE subdiscipline or choosing a discipline outside of EE, which is consistent with expectancy-value theory. These data suggest that students may pursue specific subdisciplines based on their success in the early classes. This yields another element to student motivation to choose a subdiscipline that augments the social interactions that students reported in focus groups.

Academic success, beyond GPA, may also be gauged based on participation in professional activities. Potential relationships between student achievement (GPA) and participation in

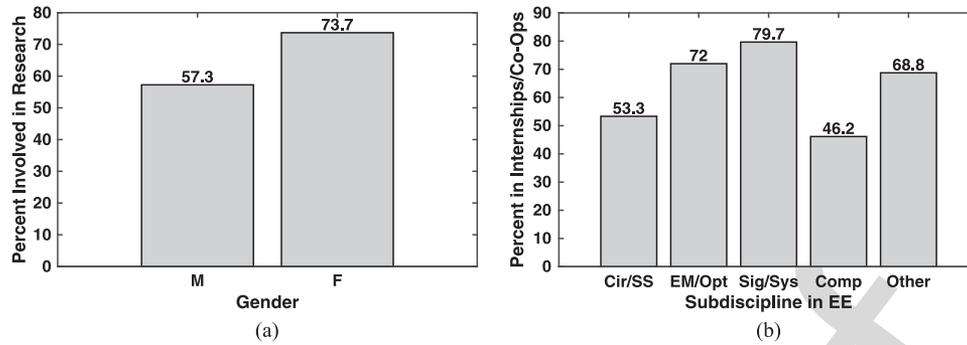


Fig. 3. Survey results for cases where association is suggested between participation in student activities and (a) gender and (b) the subdisciplines—circuits/solid state (Cir/SS), electromagnetics/optics (EM/Opt), signals and systems (Sig/Sys), and computers (Comp).

403 activities outside of the classroom did not suggest a clear asso-
 404 ciation (Table II). However, an understanding of the behavior of
 405 students across different groups may show differences related
 406 to motivations and preferences to guide academic curricula
 407 and initiatives. Comparisons between these groups based on a
 408 chi-square test for independence are summarized in Table V.
 409 Overall, the analysis does not suggest significant association
 410 between participation in student activities and gender or sub-
 411 discipline. However, there is a possible association between
 412 gender and participation in research, as well as a strong possible
 413 association between student participation in student internships
 414 subdiscipline within EE. The numerical results for these two
 415 possible associations are shown in Fig. 3. The fraction of
 416 female student participation in research is higher than for male
 417 students, a statistic that is encouraging for EE where female stu-
 418 dents are still vastly underrepresented. The fraction of students
 419 in each subdiscipline also shows significant variability, with
 420 particularly low participation for students in circuits/solid state
 421 and computers, and particularly high participation for students
 422 in the signals/systems. Further analysis is needed to determine
 423 if/how this participation relates to motivations and values for
 424 students in the various subdisciplines, and possible connection
 425 to market forces in these subdisciplines.

426

IV. CONCLUSION

427 One of the primary research objectives of this study was to
 428 examine reasons that students cite for choosing EE as a major.
 429 Based on the study, students' motivations for choosing an EE
 430 major are largely influenced by inherent interest in the subject
 431 matter and future professional opportunities. The findings for
 432 student motivation are consistent with the expectancy-value
 433 theory [8]–[11], with greatest weight on anticipated value of
 434 the professional opportunities provided by the EE degree. En-
 435 vironmental factors, including family, friends, and faculty, also
 436 play a role in student motivations, consistent with the influence
 437 of socializers on an individual's expectancies and values [9].
 438 Motivation impacts student choices, including choice of major,
 439 elective courses, and student activities, highlighting the impor-
 440 tance of examining motivations for students to choose EE and
 441 the education experience together.

442 The second related research objective of this study was to ex-
 443 amine how student performance in EE relates to demographics

and curricular choices. The data analyzed in this study indicate
 444 that student performance and participation in student activities
 445 do not appear to be associated with gender or parental educa-
 446 tional background for students in the EE program. Performance
 447 in early required courses in math and physics serve as strong
 448 predictors for future academic success, but not necessarily
 449 choice of subdiscipline in EE. In contrast, relative performance
 450 in select required courses in the EE program serves as a
 451 predictor for choice of subdiscipline, which provides further in-
 452 sight into student motivation. Participation in student activities
 453 generally does not show statistically significant association with
 454 gender or subdiscipline, with the exceptions of participation of
 455 research projects with respect to gender and participation in
 456 internships/co-ops with respect to subdiscipline. The varying
 457 correlations, or lack of correlations, between subdisciplines
 458 suggest that factors beyond grade performance play a role in
 459 student behavior. In particular, social factors may play a role
 460 in explaining these variations (consistent with the influence of
 461 socializers on expectancy-value [9]), and may strongly influ-
 462 ence student decisions with possible differences between the
 463 EE subdisciplines.
 464

The studies indicate that primary motivating factors in de-
 465 termining choice of EE major and subdiscipline are a strong
 466 perceived value of professional opportunities afforded by the
 467 technical area, and social influences, including specific faculty
 468 instructors. Early interactions with students before they declare
 469 EE or subdiscipline in EE can strongly motivate curricular and
 470 co-curricular choices and professional activities. These factors
 471 are likely to apply broadly to student populations beyond the
 472 University of Michigan. The findings suggest that it is impera-
 473 tive to connect with students early in their academic pursuits
 474 to positively influence student choices, as this may improve
 475 student recruiting and retention. Specifically at the University
 476 of Michigan, emphasis will be given to assigning the most out-
 477 standing instructors to early required courses, as active learning
 478 has been shown to increase student performance [14],[15],
 479 with anticipated increase in student motivation according to
 480 expectancy-value theory. Further recommended efforts are to
 481 expand exposure of EE-related concepts in first-year engineer-
 482 ing courses and to hold targeted informational events to show-
 483 case professional opportunities for EE majors and associated
 484 subdisciplines. Given that motivations can be largely shaped by
 485 early curricular experience, revising the curriculum to offer a
 486

487 more blended or integrated teaching of EE technical areas in
 488 early courses may provide a more optimal means of connecting
 489 with students than conventional modes of teaching topical areas
 490 in separate courses in a sequential manner. Such curricular
 491 reform may provide a means for students to make connections
 492 between technical areas while also providing a more impartial
 493 exposure to EE subdisciplines, ultimately equipping students
 494 with more accurate perceptions upon which to base future
 495 curricular and professional decisions.

496

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