



# Modeling Determinants of Undergraduate Computing Students' Participation in Internships

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# Motivation

## Industry Internships

### Students

- build technical skills
- develop professional skills
- gain authentic experiences
- explore computing pathways
- develop professional identities
- secure future employment and higher salaries

(e.g., Minnes et al., 2021)

### Employers

- provide cues for hiring full-time employees
- offer an opportunity to evaluate potential candidates

(e.g., Stepanova et al., 2021)

### Computing Programs

- increase retention in computing programs
- improve quality of capstone projects

(e.g., Fryling et al., 2018)

# Motivation



58%

Graduating computing undergraduate students who pursue an internship in US

(Kapoor and Gardner-McCune, 2020)

## Research Question

What are the factors that influence undergraduate computing students' participation in internships?

# Prior Work

- Participation in internships in computing
  - importance of internships (Kapoor and Gardner-McCune, 2019)
  - student demographics (Kapoor and Gardner-McCune, 2020)
  - barriers that support/prohibit securing internships (Kapoor and Gardner-McCune, 2020)
  - used qualitative and bivariate analysis
- Modeling internships in other domains
  - race, gender, age, first-generation status, and participation in high-impact practices such as research, etc. influenced internship participation (Hoekstra, 2021)

# Prior Work

- Theoretical background on identity
  - James Marcia's theory of identity development (Marcia, 1966)
  - four statuses across two dimensions: exploration and commitment
    - identity diffusion (**low** exploration, **low** commitment)
    - identity foreclosure (**low** exploration, **high** commitment)
    - identity moratorium (**high** exploration, **low** commitment)
    - identity achievement (**high** exploration, **high** commitment)

# Methods

- Study Design

- Larger study: mixed methods based on a Concurrent Triangulation Design
  - surveys and interviews
- This work:
  - cross-sectional survey data
  - quantitative approach to analysis

- Research question

- What are the factors that influence undergraduate computing students' participation in internships?

# Methods

- **Research Site**

- **Population:** Computing undergraduate students
- **Sample:** students enrolled in an undergraduate computing degree program at a large public university in the US
- **Context:**
  - admission to the site is selective
  - students can select a major when they start the program but have the flexibility to switch it at any time
  - students enrolled in CS, Computer Engineering (CE), and Digital Arts & Sciences (DAS) majors
  - participation in an internship is not mandatory before graduation

# Methods

- Participants and recruitment
  - study approved by the Institutional Review Board at the research site
  - participants were recruited from CS1, CS2, software engineering, HCI, and OS courses and offered extra credit or random gift cards
  - 43% response rate (N=698, Total course enrollments=1620)
  - Discarded data:
    - students who were not in CS-related majors
    - CS minors
    - students who completed less than 80% of the survey
    - students who were not in an undergraduate program
    - students without gender classification
    - non-traditional students over age 24
    - students with a high proportion of relevant missing data



# Methods

- Participants

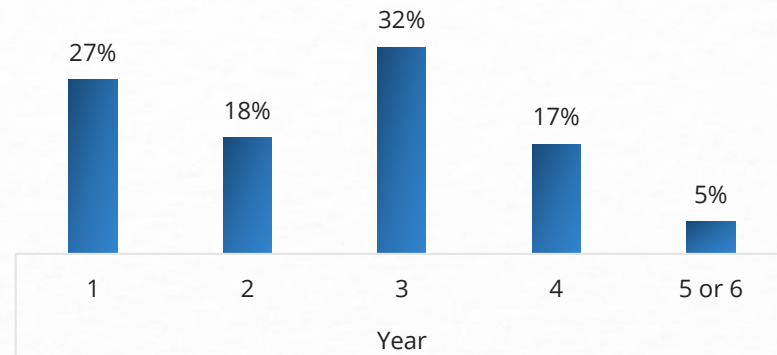
- 518 students enrolled in

- CS (66%),
    - CE (26%),
    - DAS (4%), or
    - double majors including CS (4%)

- Average age of students: 20

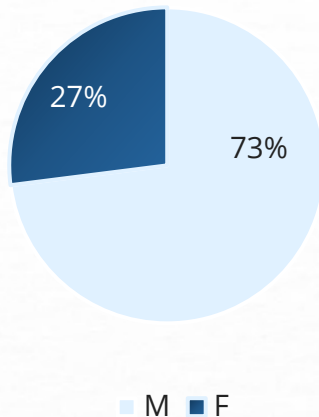
- Min = 18
    - Max = 24
    - SD = 1.4

Academic Year (N = 518)

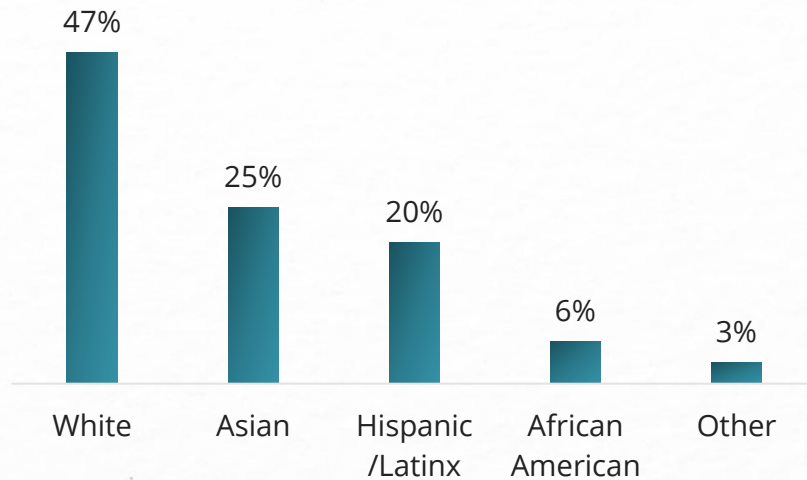


# Methods

Gender (N = 518)



Race/Ethnic Identity (N = 518)



# Methods

- Survey

- 11 sections

- demographics,
    - professional goals and identity,
    - degree program experience,
    - social support,
    - involvement in external activities

- Questions developed from

- findings of our prior work and pilot study
    - NCWIT Student Experience of the Major Survey
    - CRA Data Buddies Survey
    - Bennion and Adams' Extended Objective Measure of Ego Identity Status (EOM-EIS) instrument for measuring Marcia's identity statuses

# Methods

- Response (dependent) variable
  - binary categorical variable
  - participation in internship(s) or co-op(s) during a student's enrollment in a degree program
  - not counting internships during high school
- Explanatory or independent variables
  - 13 independent variables

# Methods

## Explanatory (independent) variable descriptions

| Variable Category                     | Independent Variable         | Description (Coded value)   |
|---------------------------------------|------------------------------|---|
| Demographic and Socioeconomic Factors | Household income ★           | {"Less than \$20,000" (1) ... "Over \$150,000" (7)}   |
|                                       | Race/ethnicity ▲             | {White/Asian (0), Underrepresented: all other ethnic and racial representations (1)}              |
|                                       | Gender ▲                     | {Male (0), Female (1)}  |
|                                       | Age ■                        | Numerical (Range: 18-24)  |
|                                       | Employment status ▲          | {Unemployed (0), Employed - working along with the degree program (1)}                            |
| Academic Profile                      | GPA ■                        | University-level grade point average on a 4.0 scale   |
|                                       | High school courses in CS ▲  | {No (0), Yes (1)}   |
|                                       | Year in school ★             | {Freshman (1), Sophomore (2), Junior (3), Senior (4), Super Senior (5)}                           |
| Identity                              | Diffusion score ☉            | Marcia identity status composite score (scale: 6-30): Low exploration, low commitment             |
|                                       | Foreclosure score ☉          | Marcia identity status composite score (scale: 6-30): Low exploration, high commitment            |
|                                       | Moratorium score ☉           | Marcia identity status composite score (scale: 6-30): High exploration, low commitment            |
|                                       | Achievement score ☉          | Marcia identity status composite score (scale: 6-30): High exploration, high commitment           |
| External Involvement                  | External involvement score ☉ | Composite score based on involvement in 14 activities, e.g., hackathons, clubs, etc.(scale: 0-42) |

Key: Binary encoded categorical ▲ | Ordinal encoded categorical ★ | Quantitative ■ | Quantitative variable computed from ordinal scale questions ☉

# Methods

- Analysis

- binary logistic regression model
  - participation in internship (Yes/No) is the response variable
  - 13 explanatory variables (excluded one variable later – multicollinearity)
- The logit (i.e., the natural logarithm of an odds ratio) forms the basis of logistic regression
- The odds ratio represents the odds that an outcome will occur (e.g., a student participates in an internship), given the presence of some factor and controlling for other predictors

# Methods

- Model

$$Z = \ln\left(\frac{P_i}{1 - P_i}\right) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_{12} X_{12}$$

- $P_i$  is the probability of event  $i$ ,
- $\beta_0$  is the constant coefficient,
- $X_1 \dots X_{12}$  are the explanatory variables,
- $\beta_1 \dots \beta_{12}$  are coefficients of explanatory variables

- Null and Alternate Hypothesis

$$H_0 = \beta_1 = \beta_2 = \dots = \beta_{12} = 0$$

$$H_A = \beta_1 = \beta_2 = \dots = \beta_{12} \neq 0$$

# Methods

- Data imputation

- handling 1.4% missing data
- 326 of the total 23828 data points were replaced

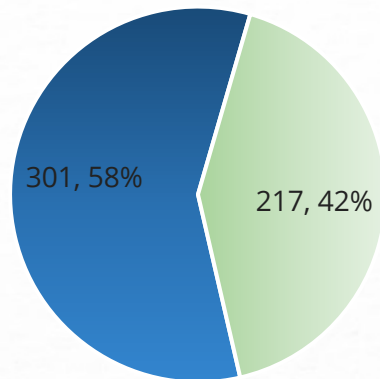
- Multicollinearity

- correlation coefficients such as Pearson's R (correlated if  $\geq \pm 0.7$ )
- variance inflation factor (correlated if  $VIF > 5$ )
- Age was highly correlated with the year in school (Pearson's R = 0.80) and hence we excluded age



# Results

Descriptive Statistics of Dependent/Response Variable (N=518)



■ No internship experience   ■ Participated in at least 1 internship

# Results

|                          | Coef. $\beta$ | Std. Err    | Z                            | $p >  Z $     | CI for Coef. $\beta$ |                      | Odds Ratio<br>$\exp(\beta)$ | CI for Odds Ratio |             |
|--------------------------|---------------|-------------|------------------------------|---------------|----------------------|----------------------|-----------------------------|-------------------|-------------|
|                          |               |             |                              |               | [0.025               | 0.975]               |                             | 5%                | 95%         |
| Const                    | -5.53         | 1.60        | -3.45                        | 0.00          | -8.67                | -2.39                | 0.00                        | 0.00              | 0.09        |
| HS CS Edu.               | 0.29          | 0.21        | 1.36                         | 0.18          | -0.13                | 0.71                 | 1.34                        | 0.88              | 2.04        |
| Employment               | 0.10          | 0.23        | 0.45                         | 0.65          | -0.35                | 0.56                 | 1.11                        | 0.70              | 1.75        |
| <b>Year in School</b>    | <b>0.57</b>   | <b>0.10</b> | <b>5.72</b>                  | <b>0.00**</b> | <b>0.38</b>          | <b>0.77</b>          | <b>1.77</b>                 | <b>1.46</b>       | <b>2.16</b> |
| GPA                      | 0.51          | 0.28        | 1.80                         | 0.07          | -0.05                | 1.06                 | 1.66                        | 0.96              | 2.89        |
| <b>Household Income</b>  | <b>0.22</b>   | <b>0.06</b> | <b>3.76</b>                  | <b>0.00**</b> | <b>0.11</b>          | <b>0.34</b>          | <b>1.25</b>                 | <b>1.11</b>       | <b>1.40</b> |
| Gender                   | -0.28         | 0.24        | -1.16                        | 0.25          | -0.76                | 0.19                 | 0.75                        | 0.47              | 1.22        |
| Race                     | 0.16          | 0.24        | 0.67                         | 0.50          | -0.31                | 0.63                 | 1.18                        | 0.73              | 1.88        |
| Moratorium Score         | 0.01          | 0.03        | 0.48                         | 0.64          | -0.04                | 0.07                 | 1.01                        | 0.96              | 1.07        |
| <b>Diffusion Score</b>   | <b>-0.06</b>  | <b>0.03</b> | <b>-2.14</b>                 | <b>0.03*</b>  | <b>-0.12</b>         | <b>-0.01</b>         | <b>0.94</b>                 | <b>0.89</b>       | <b>0.99</b> |
| Achievement Score        | 0.01          | 0.03        | 0.29                         | 0.77          | -0.05                | 0.07                 | 1.01                        | 0.95              | 1.07        |
| Foreclosure Score        | -0.01         | 0.03        | -0.34                        | 0.74          | -0.06                | 0.05                 | 0.99                        | 0.94              | 1.05        |
| <b>Involvement Score</b> | <b>0.12</b>   | <b>0.02</b> | <b>6.79</b>                  | <b>0.00**</b> | <b>0.09</b>          | <b>0.16</b>          | <b>1.13</b>                 | <b>1.09</b>       | <b>1.17</b> |
| No. of Observations: 518 |               |             | Pseudo R <sup>2</sup> : 0.21 |               |                      | LLR p-value: 4.7e-25 |                             |                   |             |
| Df Residuals: 505        |               |             | Log-Likelihood: -279.5       |               |                      | * $p < 0.05$         |                             |                   |             |
| Df Model: 12             |               |             | LL-Null: -352.2              |               |                      | ** $p < 0.001$       |                             |                   |             |

# Results

- Model fit evaluation:
  - McFadden's pseudo- $R^2$  coefficient ( $\rho^2$ ) = 0.21
  - indicates an excellent model fit
  - According to McFadden, "values of .2 to .4 for  $\rho^2$  represent an excellent fit"

# Discussion

- Participation in internships is associated with:
  - year in school
  - household income
  - external involvement score
  - diffusion identity score (low exploration and low commitment)
- Compared with previous work on undergraduate internship participation (Hoekstra, 2021)
  - Similar results:
    - year in school
    - external involvement
  - New results:
    - race and gender were not associated with internship participation
    - household income and diffusion identity score predict internship participation

# Conclusion

- We recommend CS departments :
  - provide resources and opportunities for students' participation in activities outside the curriculum by funding student organizations, coding competitions, etc.
  - underscore the importance of involvement in external activities
  - develop support programs for students from low socio-economic backgrounds to prepare for internship recruitment process

# Limitations

- EOM-EIS identity scales had lower internal consistency due to the limited number of items used for each status
- our findings could be biased by the choice of our modeling technique
- observational study and results should not be interpreted as causal relationships
- data is from a modest sample of computing undergraduates enrolled at a single institution in the US where participation in internships was optional

# Questions



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Thanks to the SIGCSE Special Project Grant for  
supporting our project 😊