Exploring Awarding Gaps at UK Higher Education Institutions

First Year Presentation – Physics Education Research

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- 1. Introduction & Background
- 2. Methodology
- 3. Results
- 4. Conclusions

Definition

Differential degree outcomes between different demographic groups

- UK studies started at the turn of the 21st century [Smith and Naylor, 2001]
- More recent studies tend to foreground sociodemographic factors, such as ethnicity [Codiroli Mcmaster, 2021] and socioeconomic classification [Boero et al., 2022]
- Increased obstacles in higher education (HE) lead to longitudinal under-representation, widening the opportunity gap



Figure 1: Example hierarchical structure of students grouped into HE providers

- Common use of data from multiple contexts for higher statistical power
- Grouping (or 'hierarchical') structure (Fig 1) is unaccounted for - model structure is not correctly identified, weakening parameter estimates
- Hierarchical logistic regression (HLR) can account for multilevel structures [Van Dusen and Nissen, 2019]

Large HESA dataset to study degree outcomes of UK physics graduates:

HESA data

- 30,185 physics and astrophysics graduates from IOP-accredited Honours degrees
- 2013/14 2021/22
- A-Level or equivalent qualifications only
- Dependent variable: Binary did / did not achieve First Class Honours
- Independent variables:
 - Academic information: UCAS Tariff point score, course duration and provider
 - $\bullet\,$ Sociodemographic information: disability status, sex, ethnicity, nationality

- Building logistic regression models of increasing complexity
- Model differing odds of achieving a First between groups of students in the data
- Utilise large dataset and hierarchical modelling (students grouped into providers)

Research questions

- 1. What are the predictors for achieving a First in Physics?
- 2. Do awarding gaps exist in Physics at UK HE institutions?
- 3. Does chosen methodology affect the size of observed awarding gaps?

Logistic regression (LR) models

Univariable (ULR)



Multivariable (MLR)

Hierarchical (HLR)



Figure 2: Visualisation of predictor effect size in ULR

Figure 3: Visualisation of predictors' effect sizes in MLR

Figure 4: Visualisation of predictors' effect sizes & random intercepts in HLR

• Odds ratio: Magnitude difference in odds between group n and 'reference' group

RQ3 - comparing models



Figure 5: Difference in age odds ratios (relative to students aged 21-24) based on chosen methodology, with 95%confidence intervals

Figure 6: Difference in school type odds ratios (relative to state-educated students) based on chosen methodology, with 95% confidence intervals

Unknown

$\mathrm{RQ1}\&2$ - predictors & awarding gaps



Figure 7: Forest plot of odds ratios for HLR-predicted significant model predictors, relative to reference levels, with 95% confidence intervals

RQ1&2 - predictors & awarding gaps



Figure 8: Predicted baseline probability \hat{p}_j at each provider in the data, based on random intercepts μ_{0j} , for a student belonging to the reference level of every fixed effect - 95% confidence intervals shown

RQ2 - predicted probabilities





Figure 9: Predicted probability of achieving a First across UCAS tariff scores for students of different ethnicities - based on MLR model, with 95% confidence intervals

Figure 10: Predicted probability of achieving a First across UCAS tariff scores for students of different ethnicities - based on HLR model, with 95% confidence intervals

Conclusions & next steps

Research questions:

- 1. **Predictors** HE provider, year, UCAS tariff points, sex, school type, nationality, ethnicity, degree duration, disability status, course type, graduating age
- 2. Awarding gaps male students, non-EU students, students with ethnicity 'Asian', 'Black', or 'Other', students with a declared disability
- 3. **Methodology** significant clustering observed

Next steps

- Intersectionality
- Longitudinal analysis
- Methodology
- Intervention

Thank you!

References

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