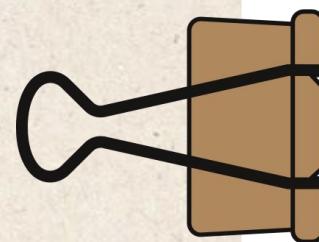


MBA Admissions

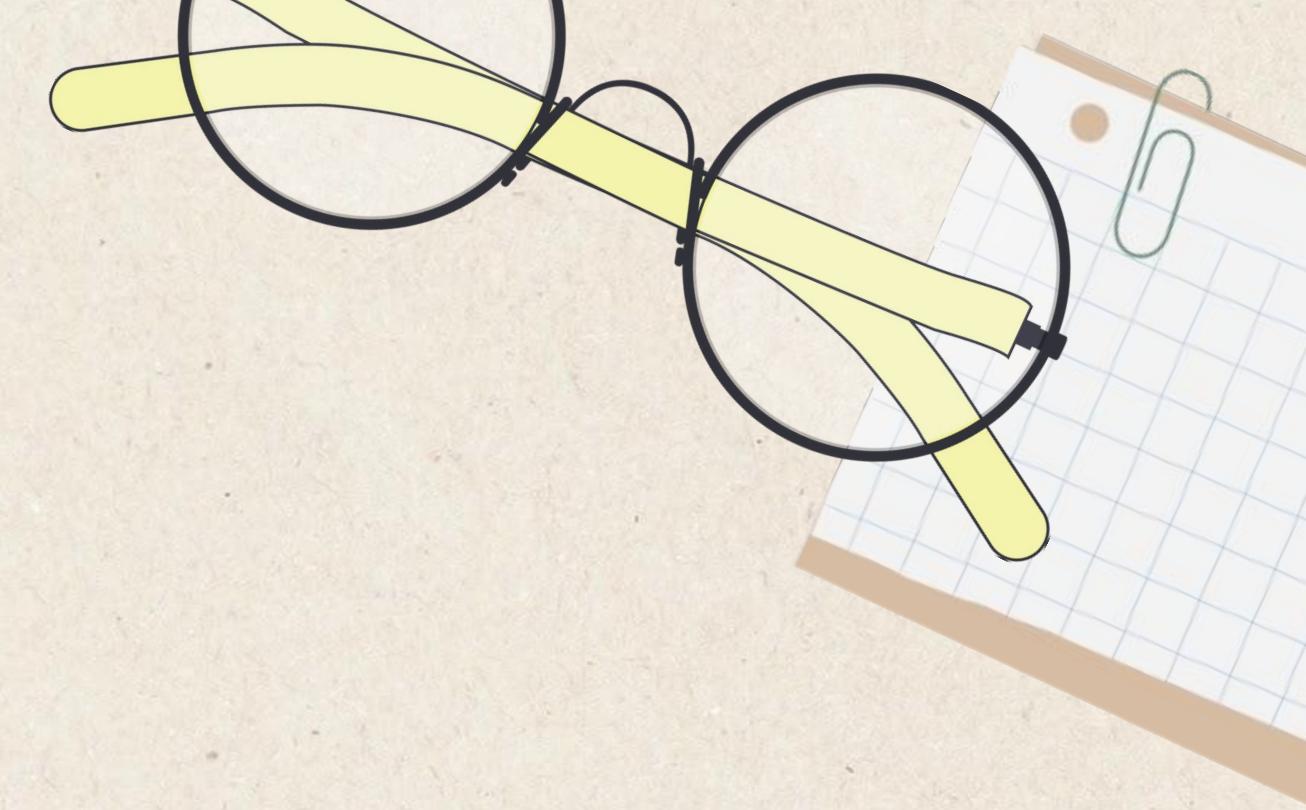
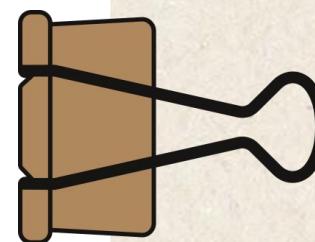
Mary Behnke

AGENDA



- Dataset Information
- Question
- Preparation
- Exploratory Data Analysis

- Rejected vs Not Rejected
- Admitted vs Waitlisted
- Gender & International
- Takeaways



DATASET



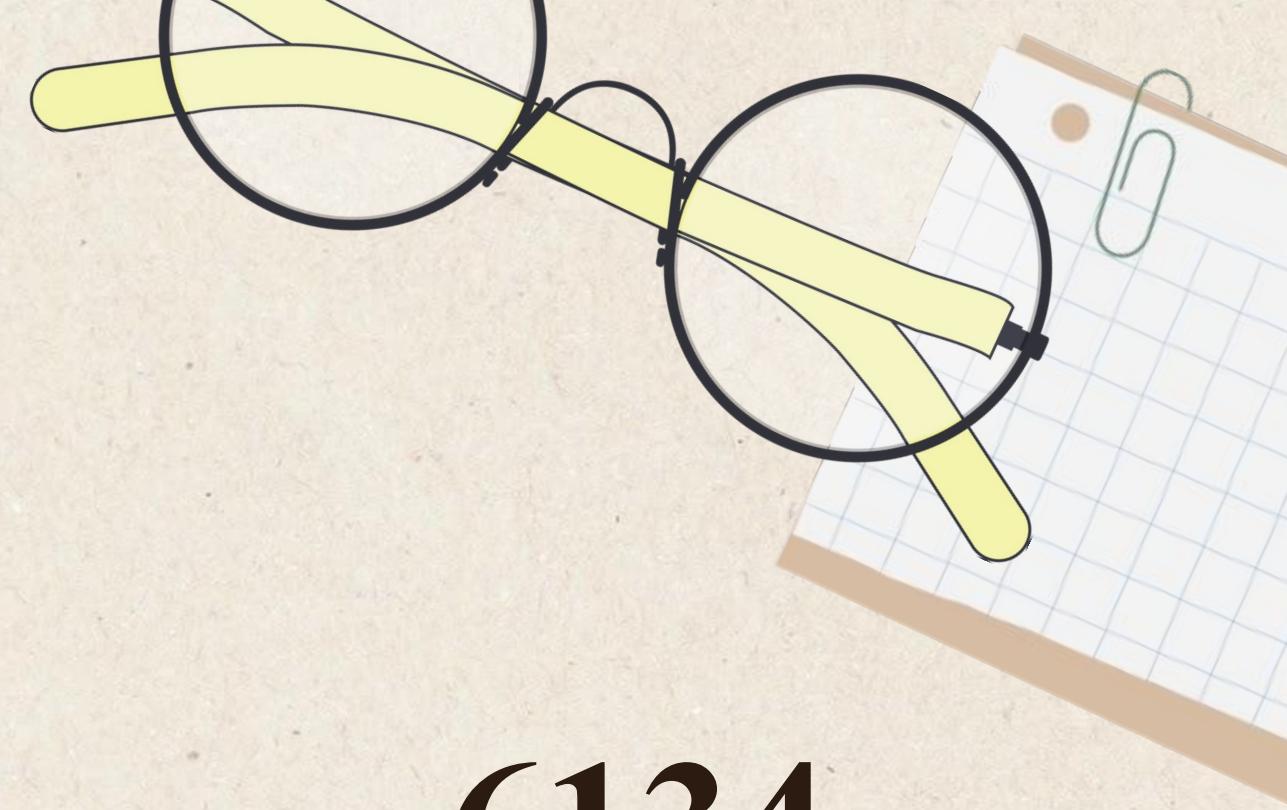
10

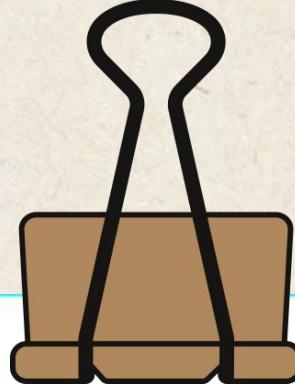
Variables

*applicationID, gender, international, gpa, major,
race, gmat, work_exp, work_industry, admission*

6134

Observations





Can a model predict whether an applicant will be rejected or admitted?

Is there any evidence for bias in admissions?

PREPARATION



Dealing with NA Values

- *race* column
- imputed “Not Provided”
- *admission* column
- imputed “Rejected”



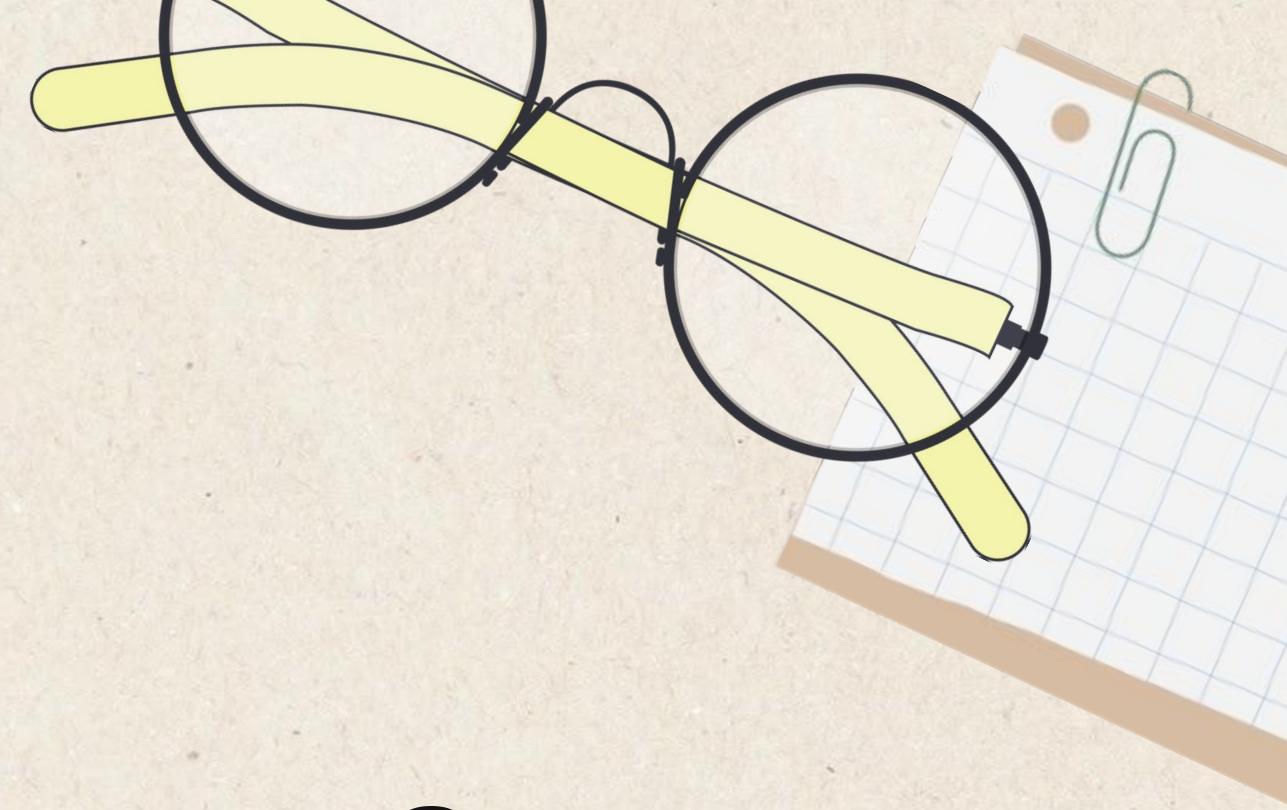
Creating New Columns

- Four new columns (all factors):
 - *rejectID*
 - *admitID*
 - *genderID*
 - *intID*



Characters to Factors

- *gender, international, major, work_industry, admission, and race*



EDA

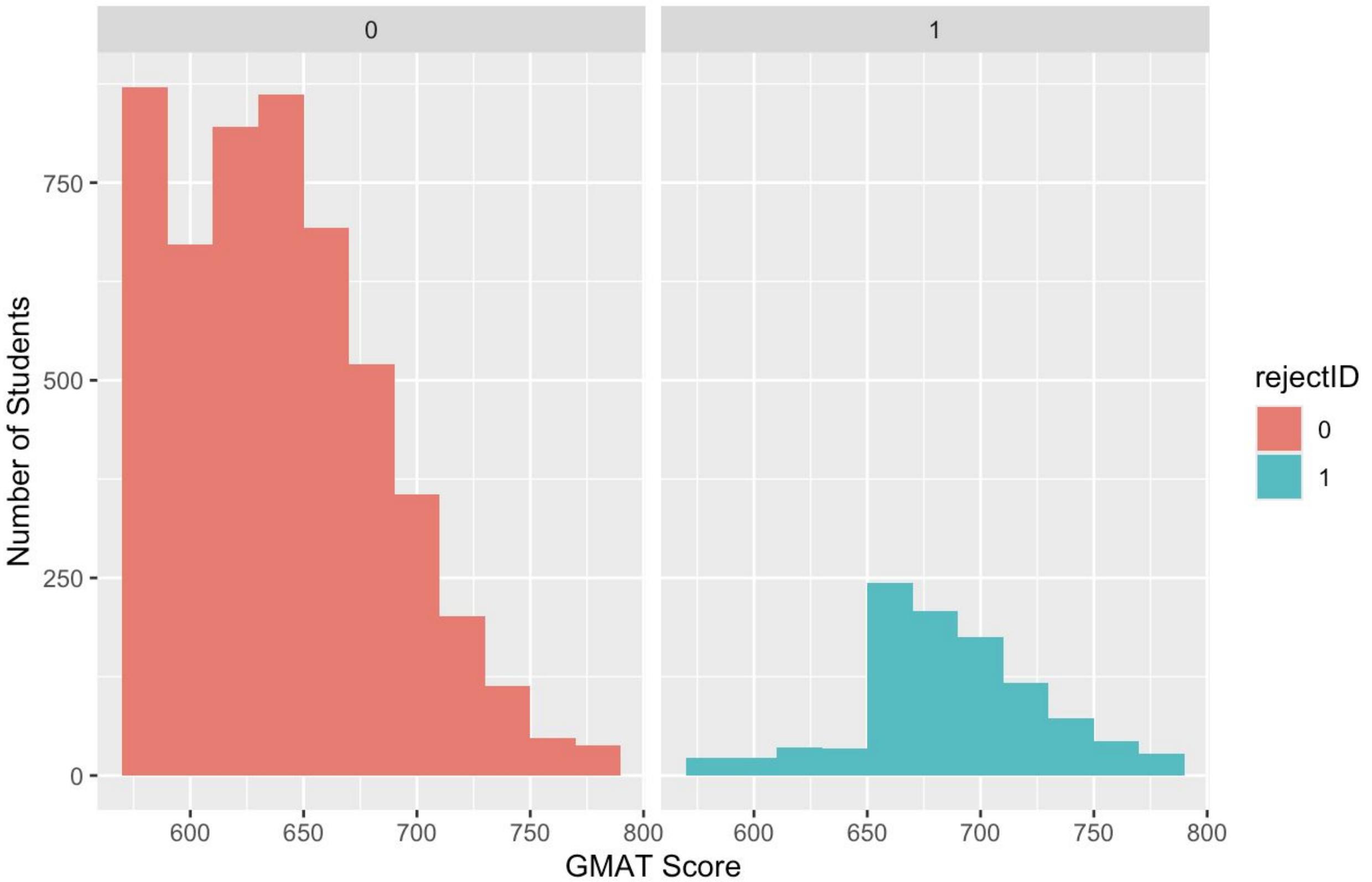
```
##   application_id    gender    international      gpa      major
## Min. : 1 Female:2251 False:4352 Min. :2.650 Business :1838
## 1st Qu.:1549 Male :3943 True :1842 1st Qu.:3.150 Humanities:2481
## Median :3098          Median :3.250 STEM :1875
## Mean   :3098          Mean  :3.251
## 3rd Qu.:4646          3rd Qu.:3.350
## Max.   :6194          Max.  :3.770

##
##           race          gmat      work_exp
## Asian      :1147 Min.   :570.0 Min.   :1.000
## Black      : 916 1st Qu.:610.0 1st Qu.:4.000
## Hispanic   : 596 Median :650.0 Median :5.000
## Not Provided:1842 Mean   :651.1 Mean   :5.017
## Other      : 237 3rd Qu.:680.0 3rd Qu.:6.000
## White      :1456 Max.   :780.0 Max.   :9.000

##
##           work_industry admission
## Consulting                  :1619  Admit   : 900
## Private Equity/Venture Capital: 907 Rejected:5194
## Technology                 : 716  Waitlist: 100
## Nonprofit/Gov              : 651
## Investment Banking           : 580
## Financial Services           : 451
## (Other)                     :1270
```

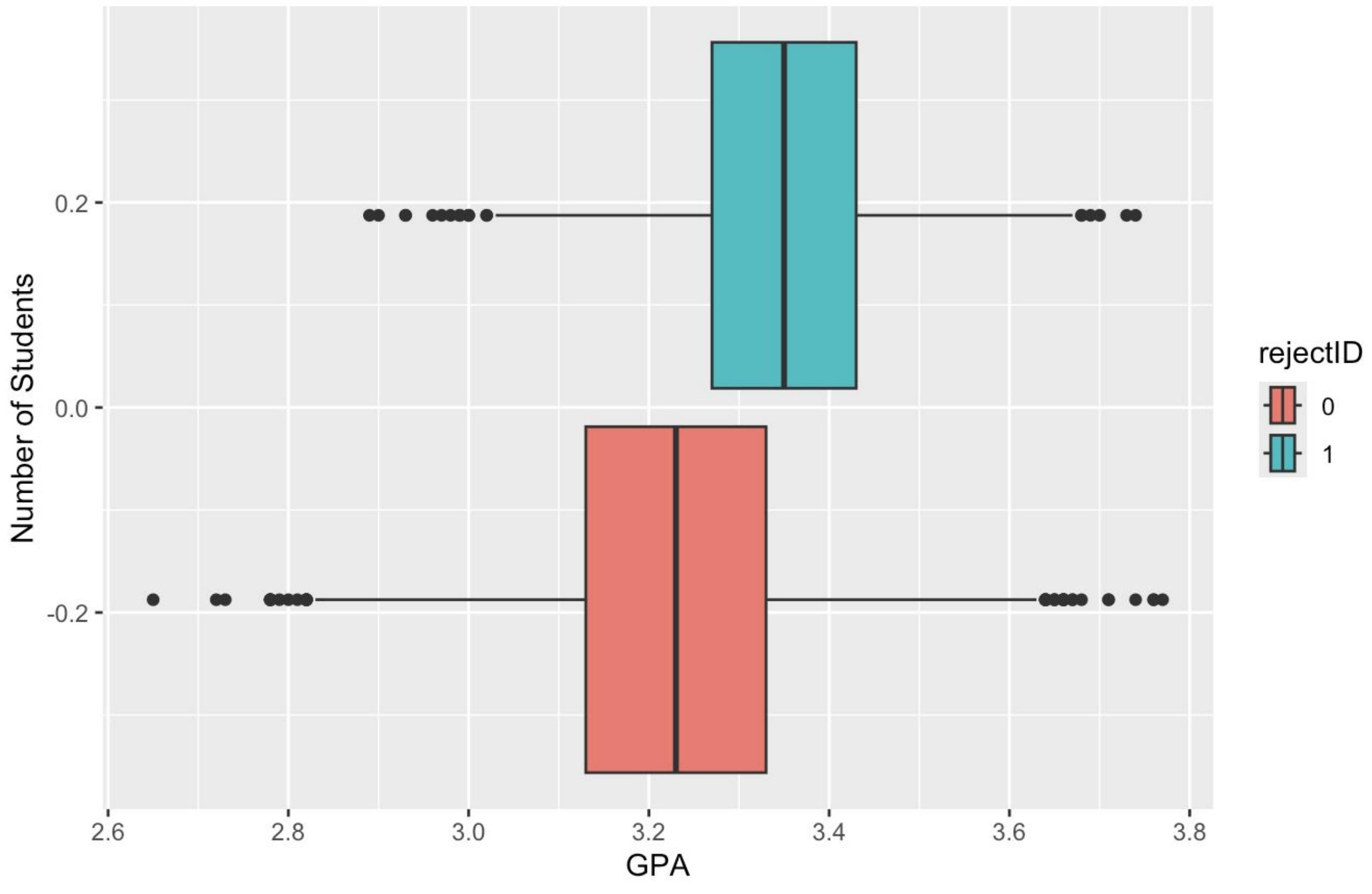
EDA

Distribution of GMAT Scores By Rejected or Not Rejected



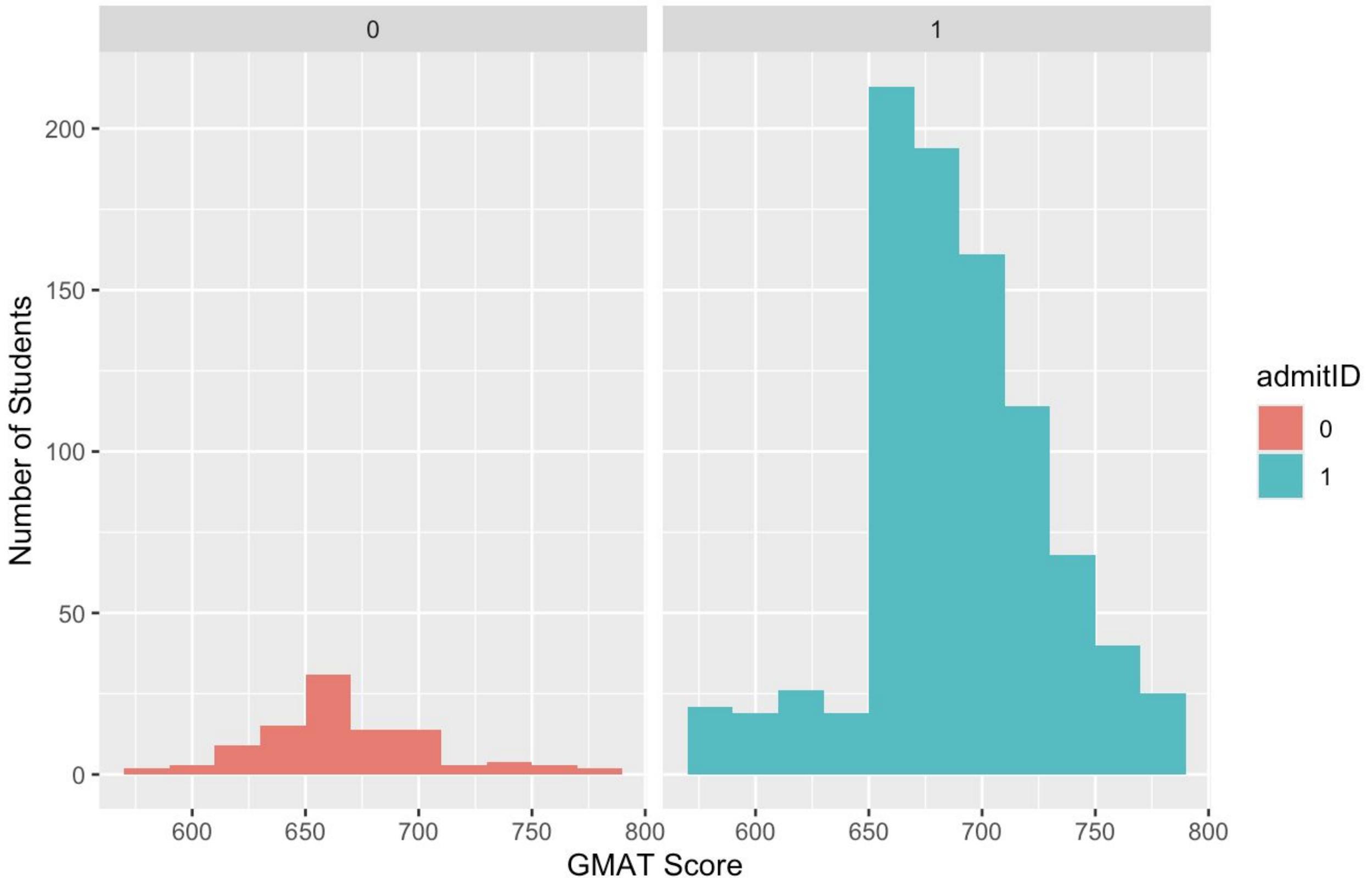
EDA

Distribution of GPA
By Rejected or Not Rejected



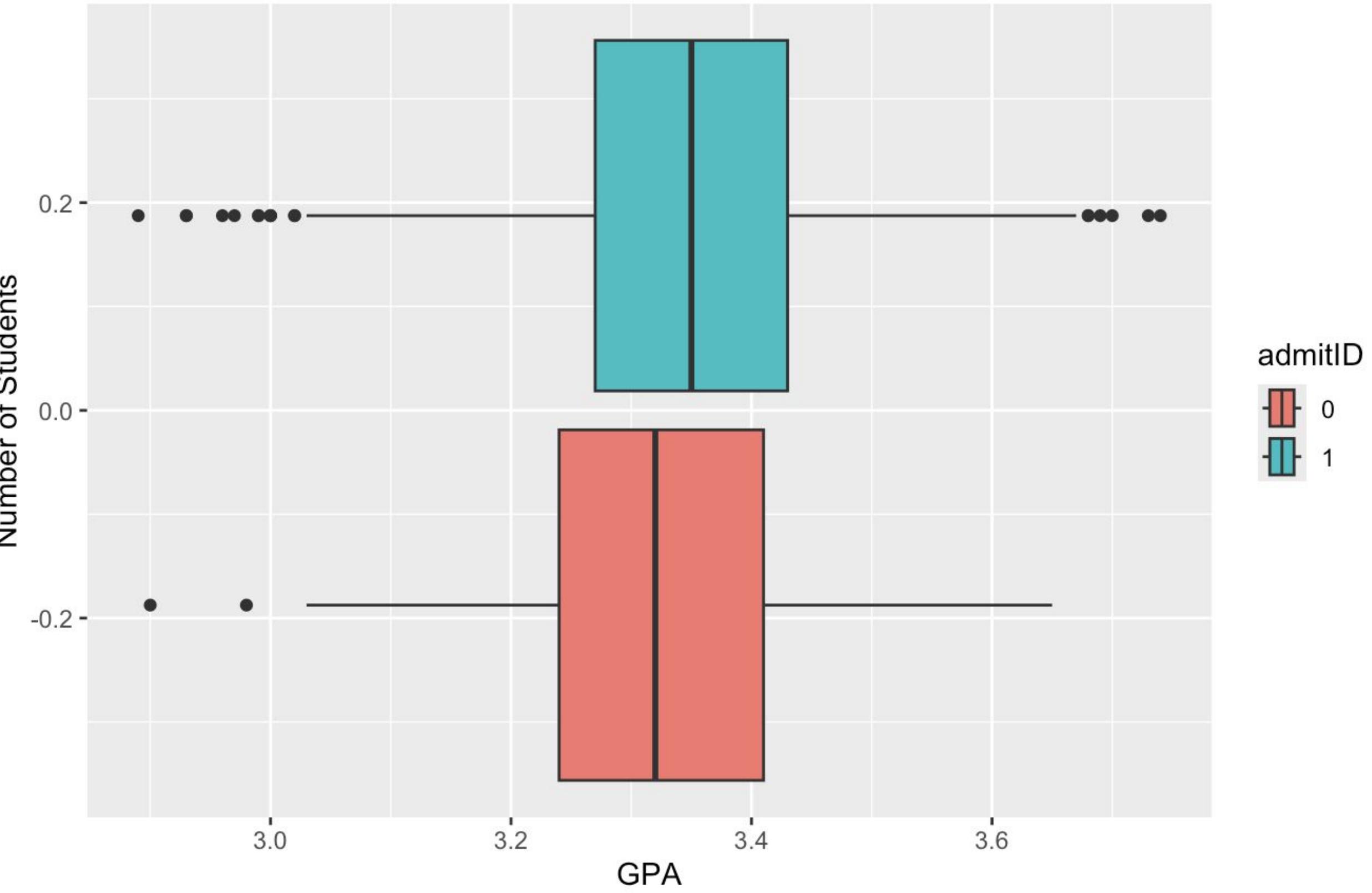
EDA

Distribution of GMAT Scores By Admitted or Waitlisted



EDA

Distribution of GPA
By Admitted or Waitlisted



REJECTED



Logistic Regression

Training set: 70%

Validation set: 30%

Predictors: *gender, international, gpa, major, race, gmat, work_exp, work_industry, gpa:race*

Significant predictors:

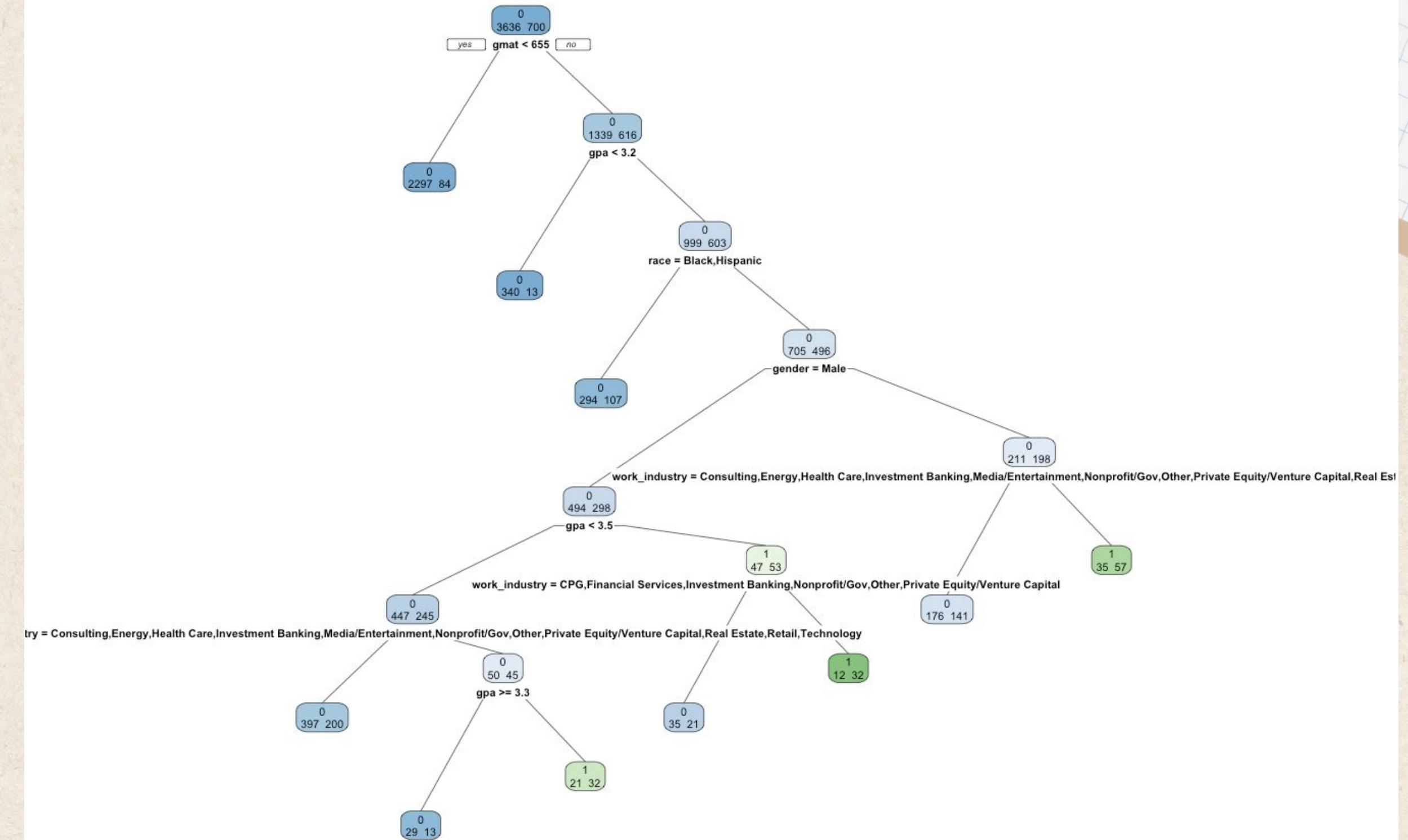
- (***) *genderMale, gmat*
- (*) *raceOther, work_industryInvestment*

Management, gpa:raceOther

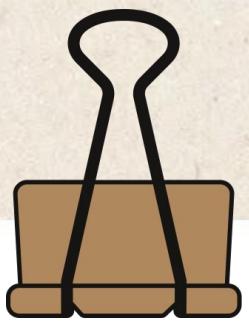
83.85%
Accuracy

Decision Tree

84.18%
Accuracy



ADMITTED



Logistic Regression

Training set: 70%

Validation set: 30%

Predictors: *gender, international, gpa, major, race, gmat, work_exp, work_industry, gpa:race*

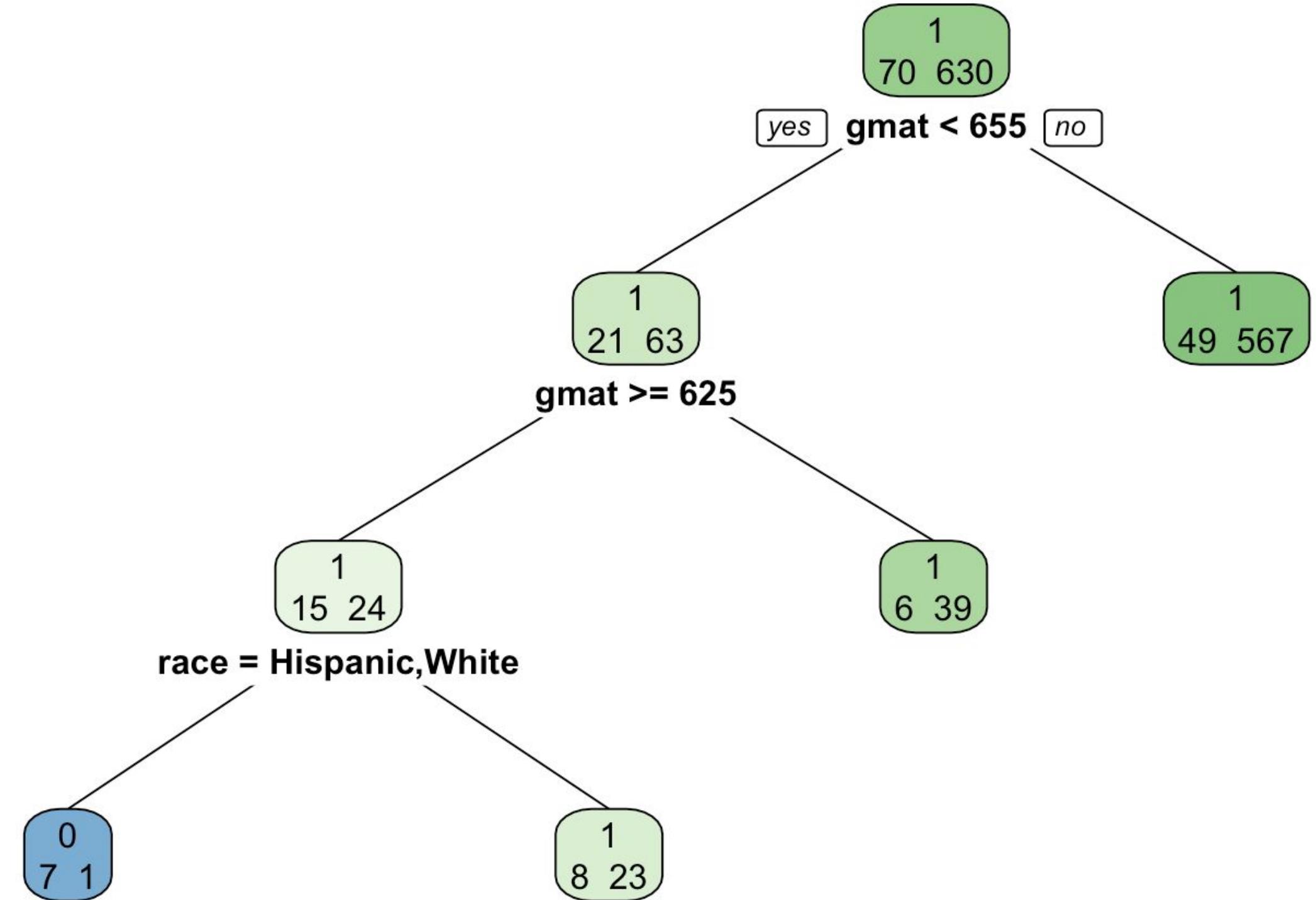
Significant predictors:

- (**) *gmat*
- (*) *genderMale, work_industryFinancial Services, work_industryTechnology*
- (.) *majorStem, work_industryInvestment Banking, work_industryInvestment Management, work_industryNonprofit/Gov*

88%
Accuracy

Decision Tree

90.33%
Accuracy



UPLIFT MODELS



Gender

Rejected vs Not Rejected

- With all observations set to *male*, there was an average -10% in prediction for not rejected

Admitted vs Waitlisted

- With all observations set to *male*, there was an average -5% in prediction for admitted



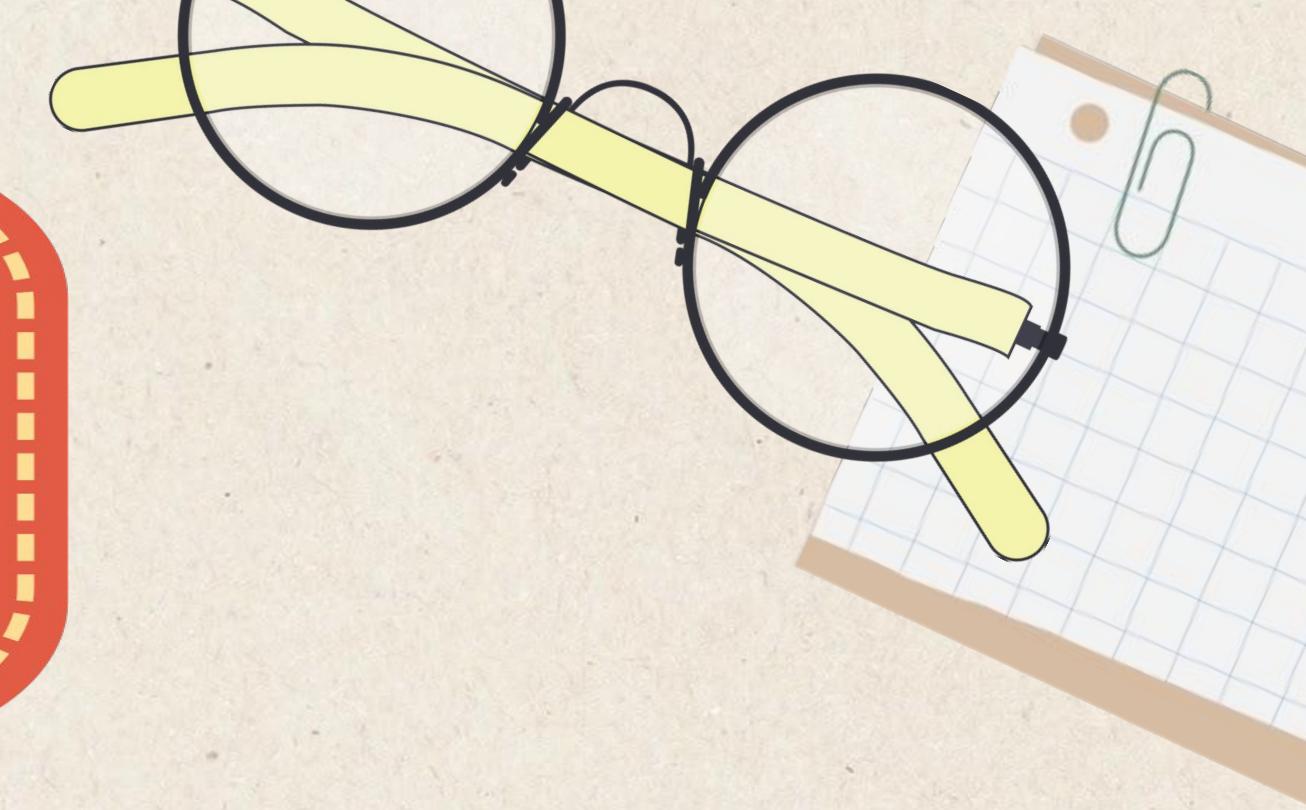
International

Rejected vs Not Rejected

- With all observations set to *international*, there was an average -8% in prediction for not rejected

Admitted vs Waitlisted

- With all observations set to *international*, there was an average +31% in prediction for admitted



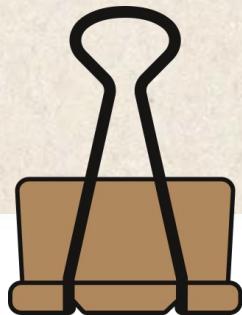
TAKEAWAYS



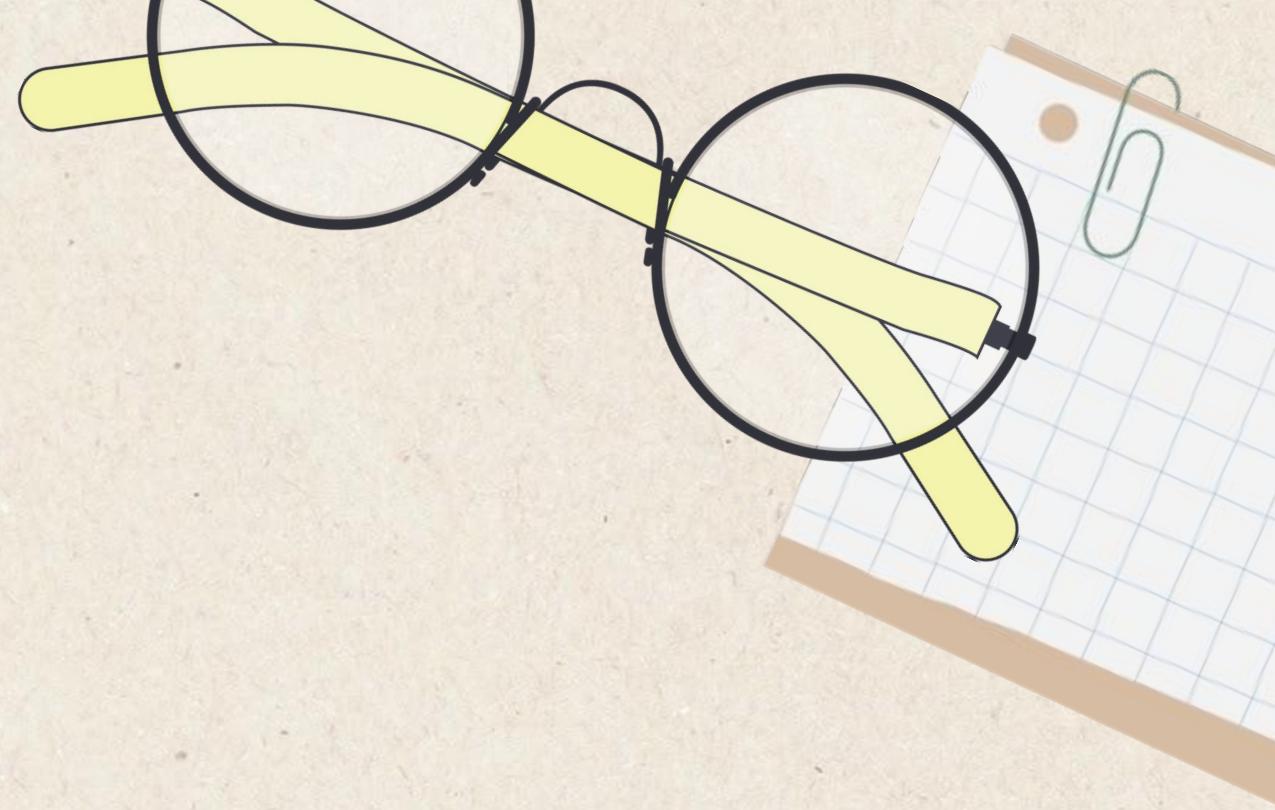
Relatively
accurate
models can be
created



Possible bias
for gender and
international
status



Could later
look more into
work industry
and race factors



thank
you!

Any Questions?

