

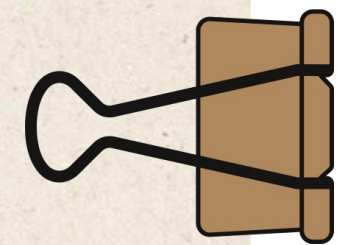


MBA Admissions

Mary Behnke

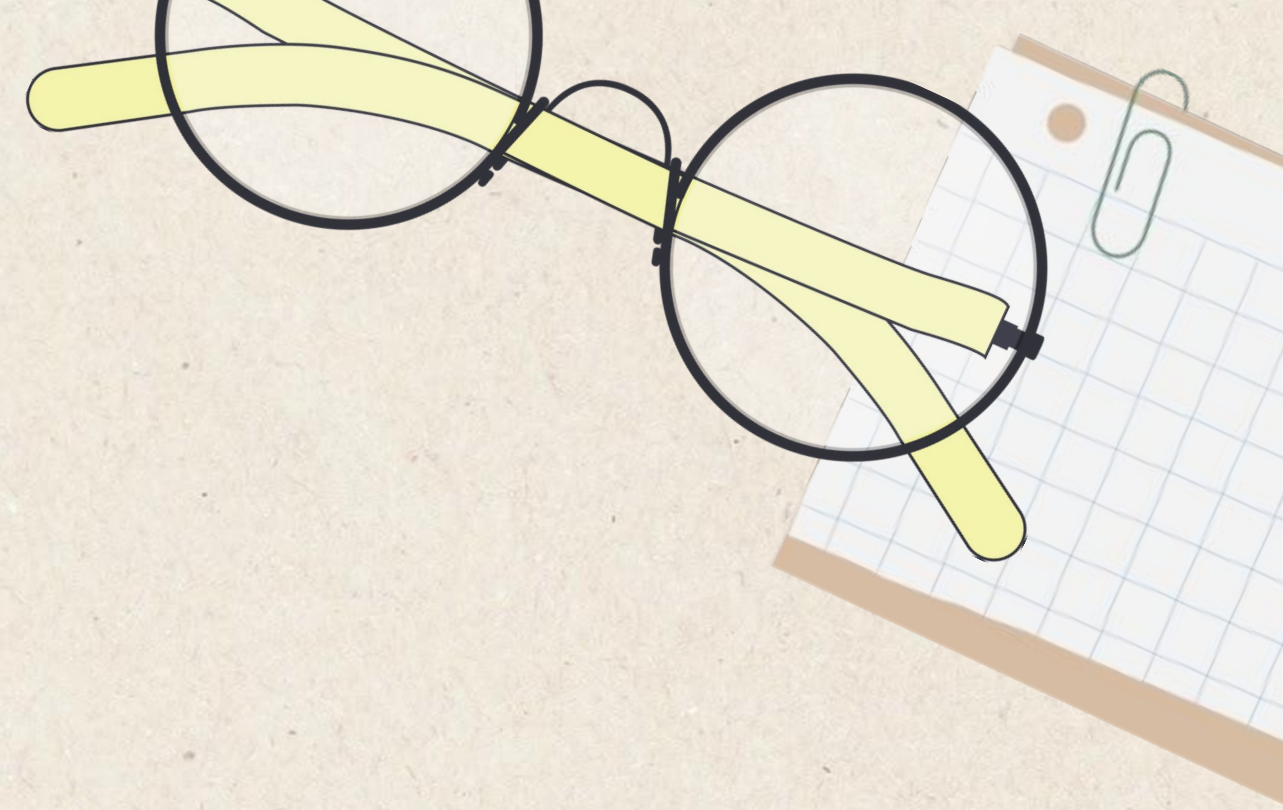
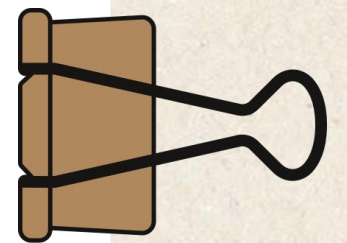
Predictive Analytics | 2024

AGENDA

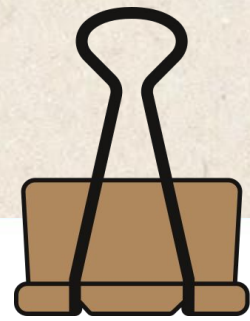


- Dataset Information
- Question
- Preparation
- Exploratory Data Analysis

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



DATASET



MBA Dataset

source: Kaggle

10


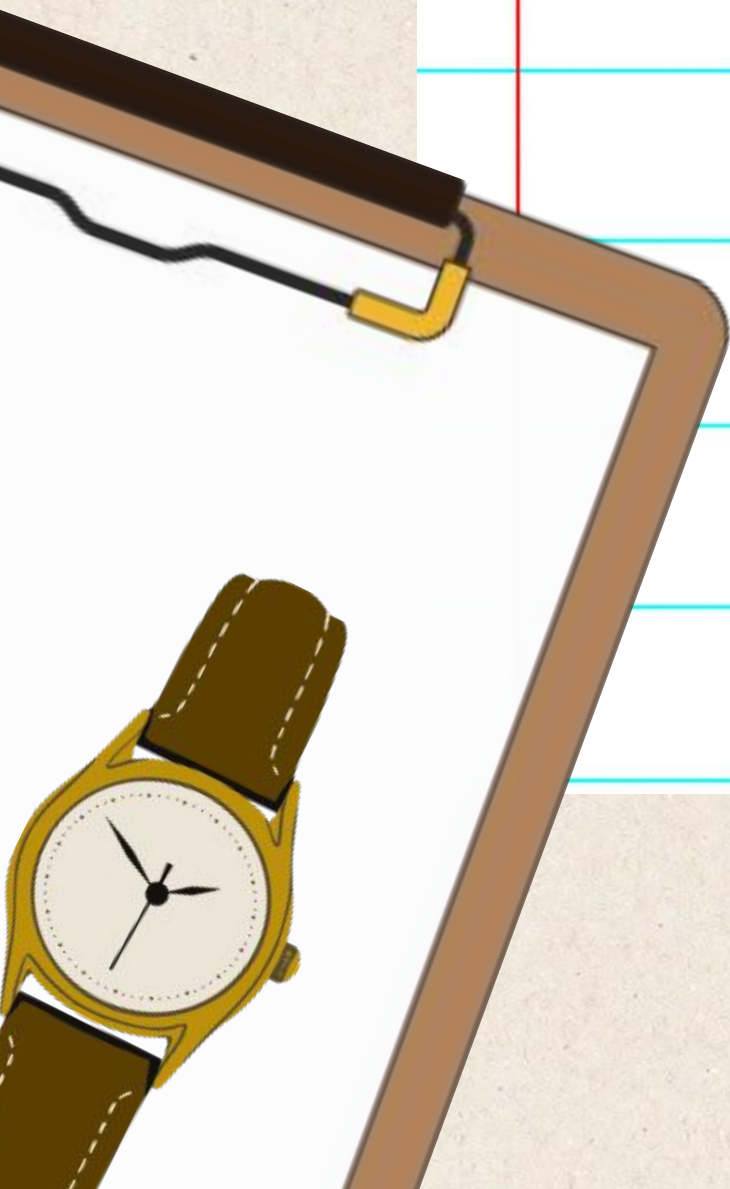
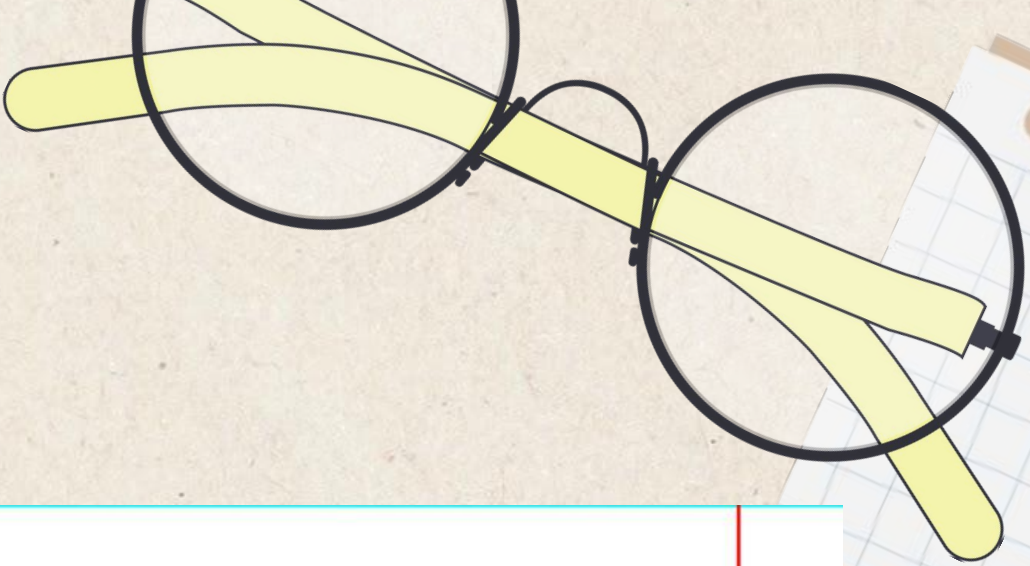
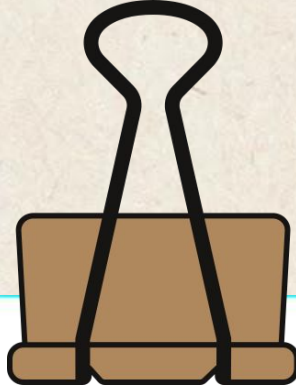
Variables

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

6134

Observations

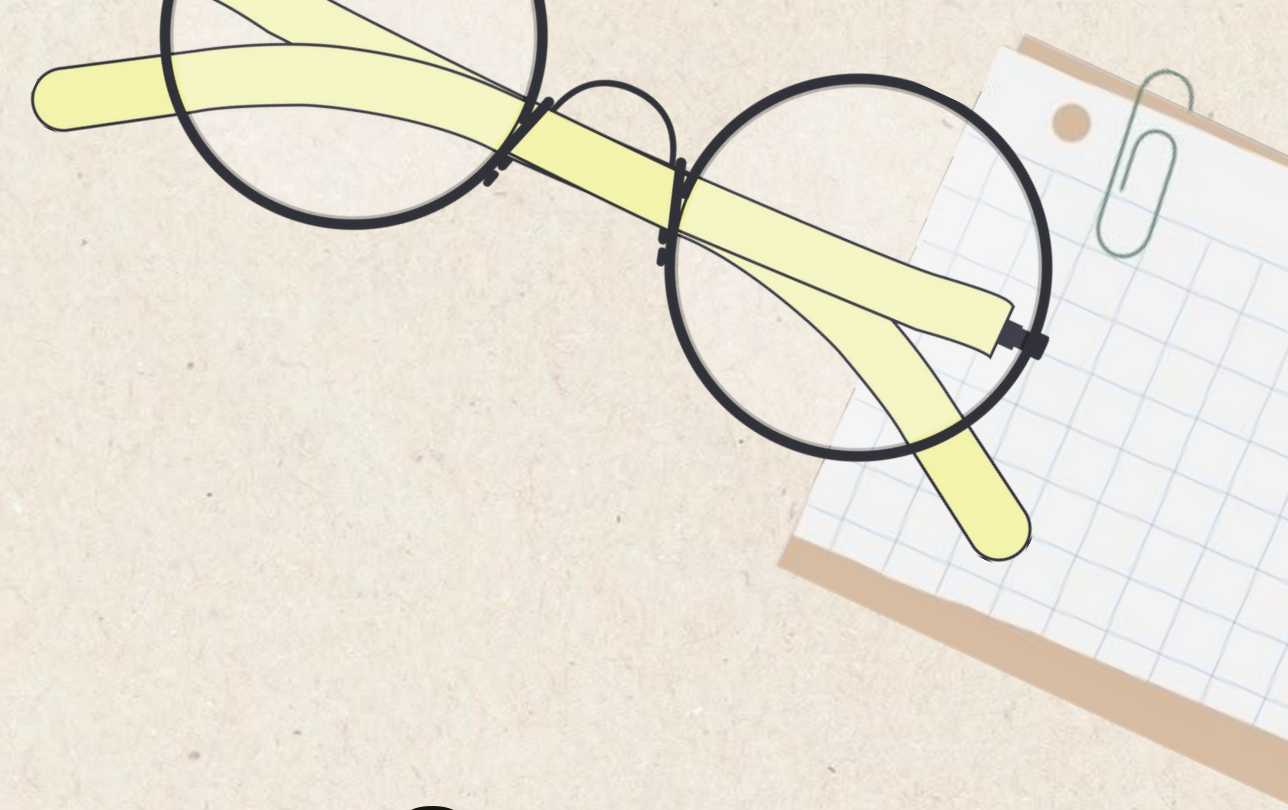




Can a model predict whether an
applicant will be rejected or
admitted?

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*



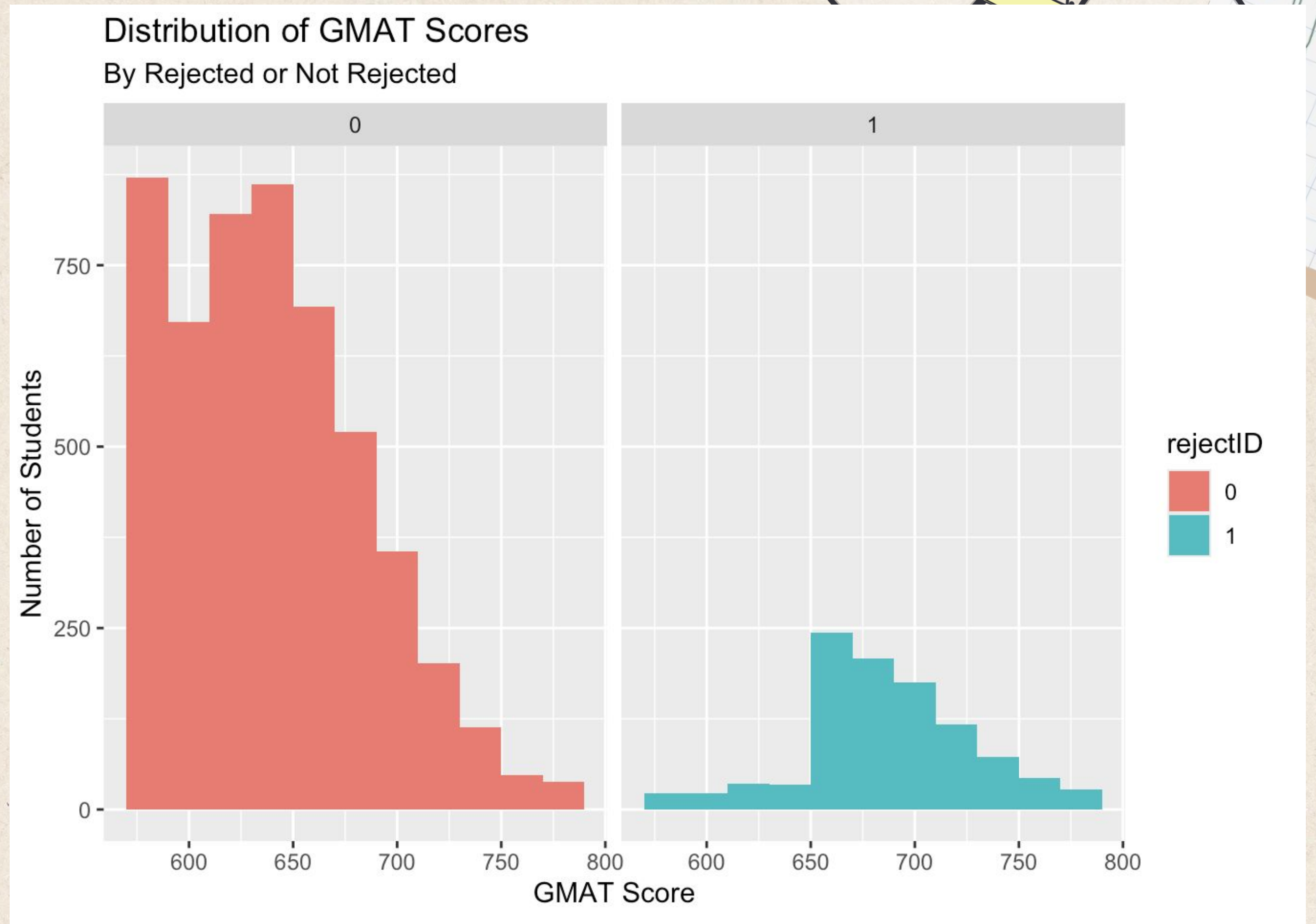
Characters to Factors

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

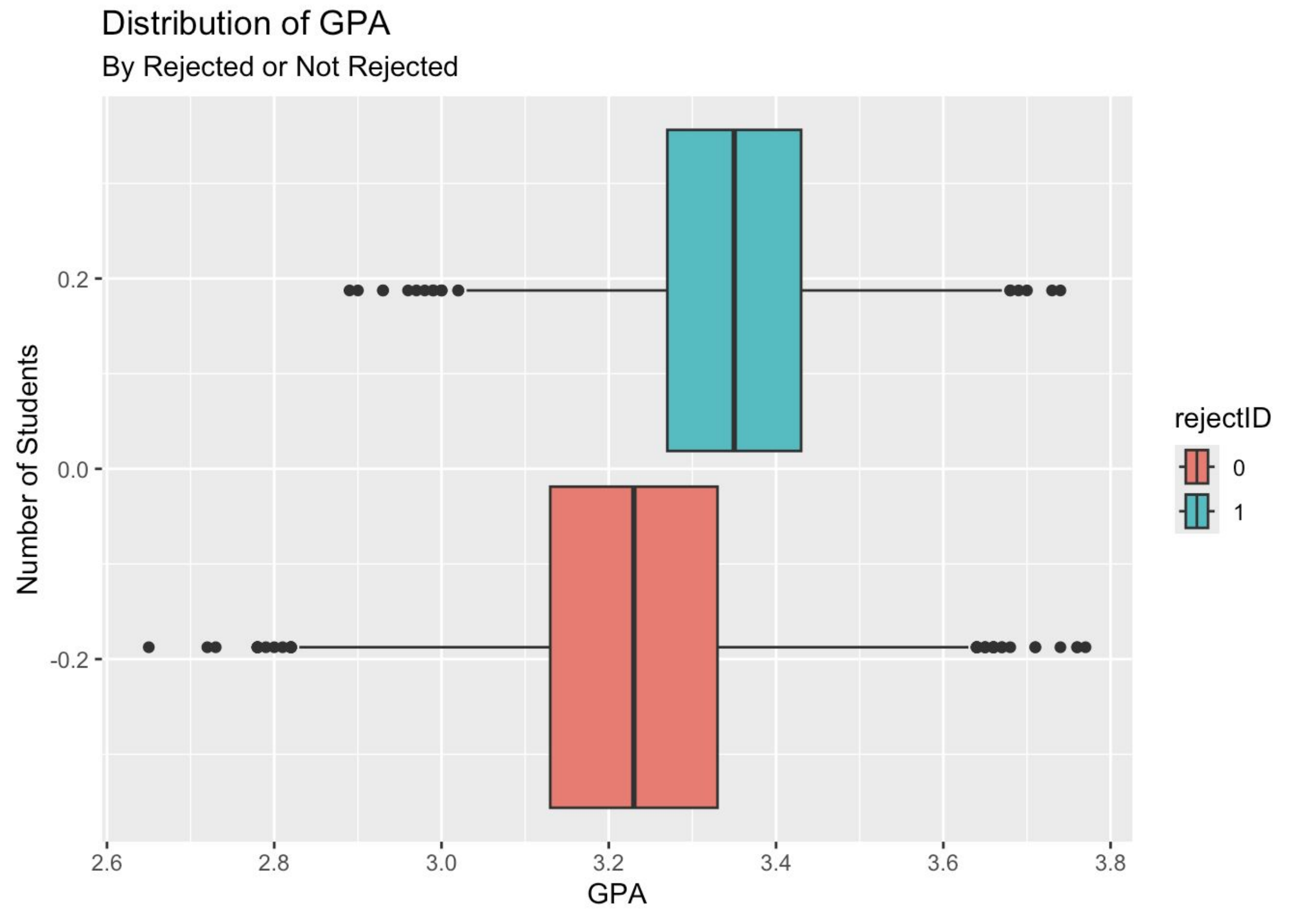
EDA

```
## 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
## Mean      :3098
## 3rd Qu.:4646
## Max.      :6194
##
##              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

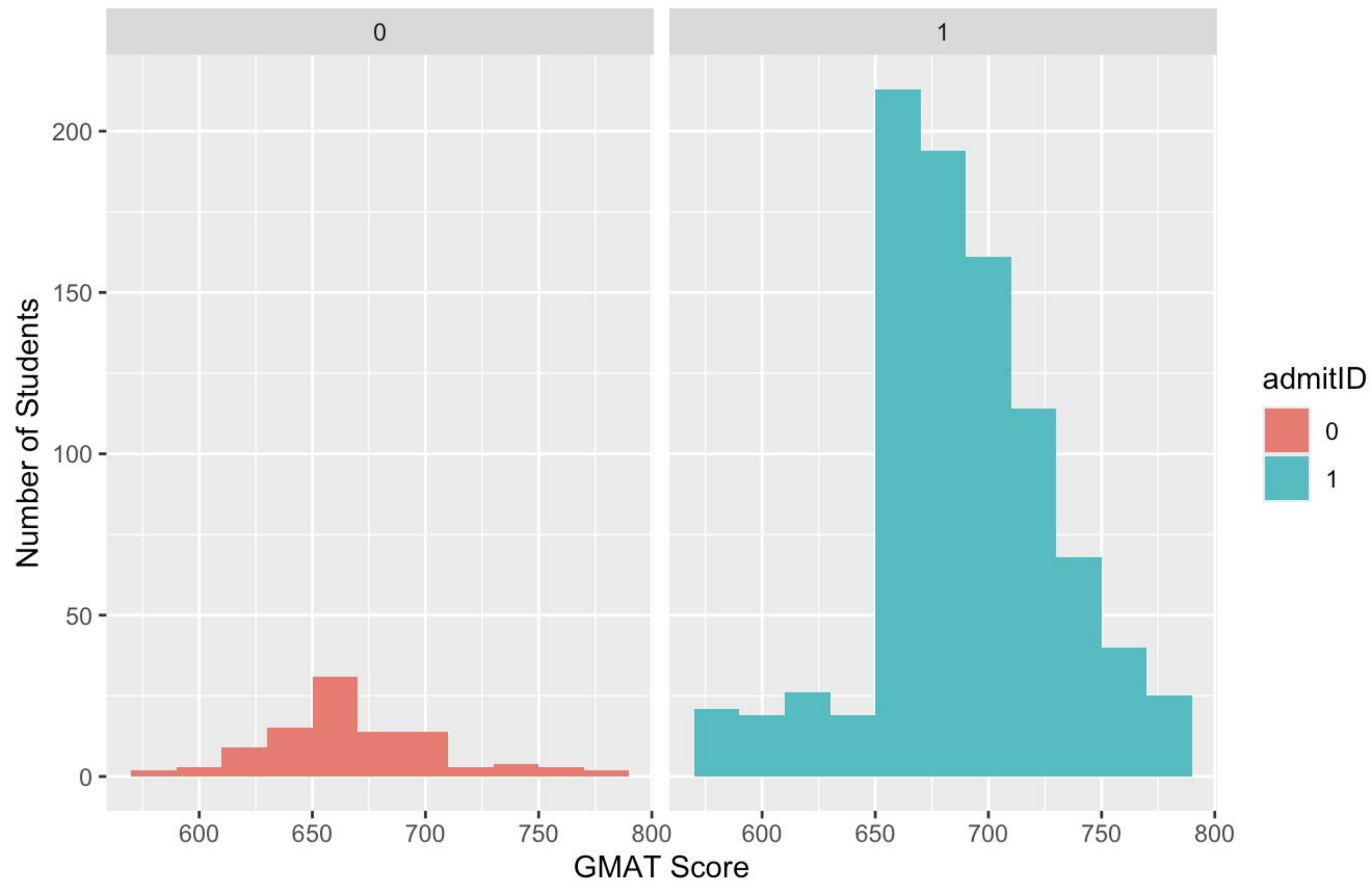


EDA

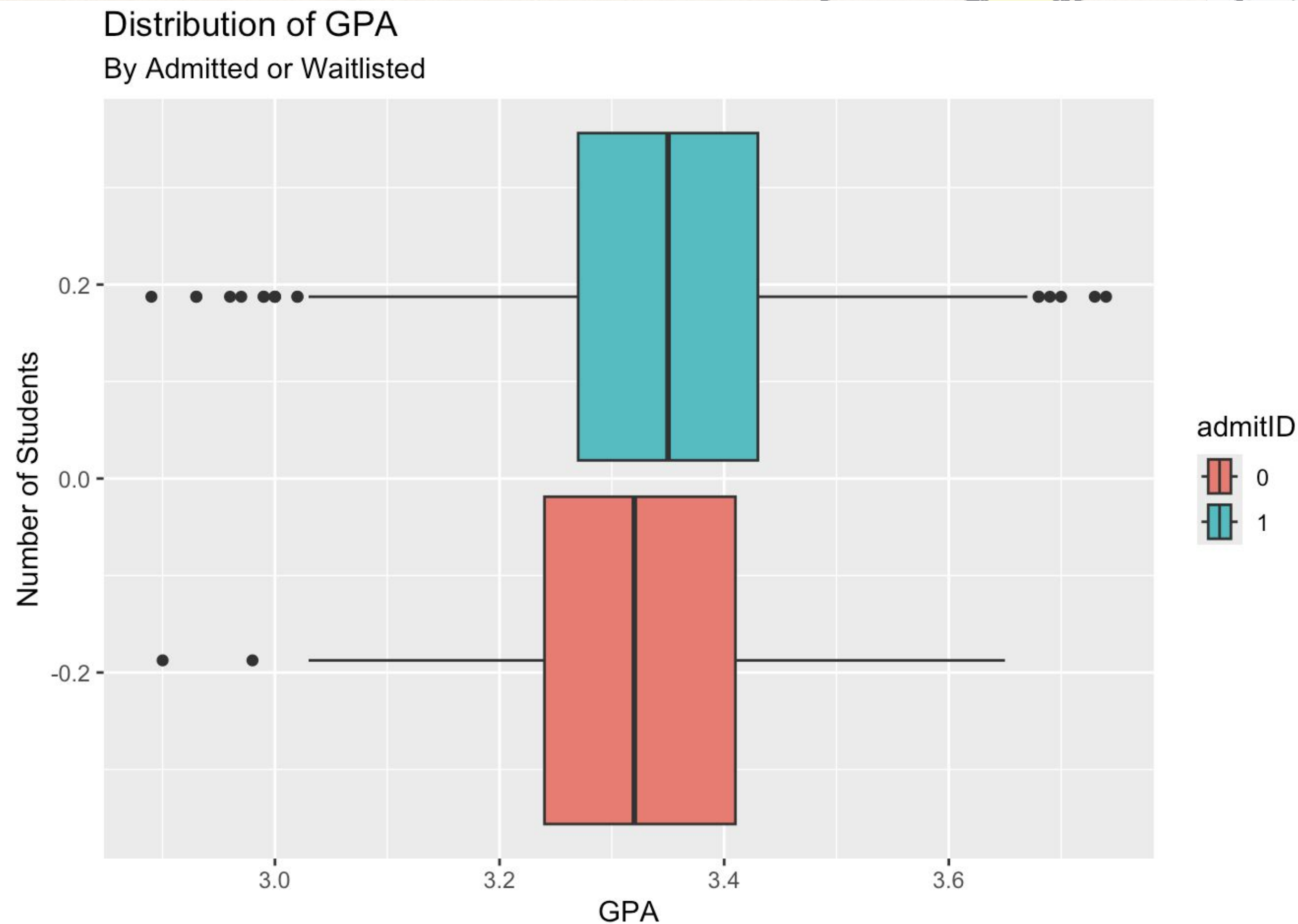


EDA

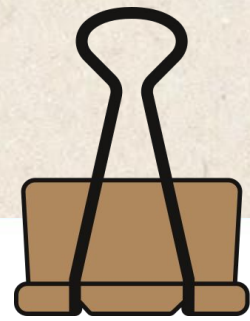
Distribution of GMAT Scores
By Admitted or Waitlisted



EDA



REJECTED



Logistic Regression



83.85%
Accuracy

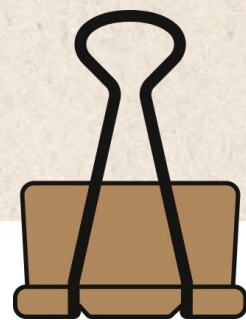
Training set: 70%

Validation set: 30%

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

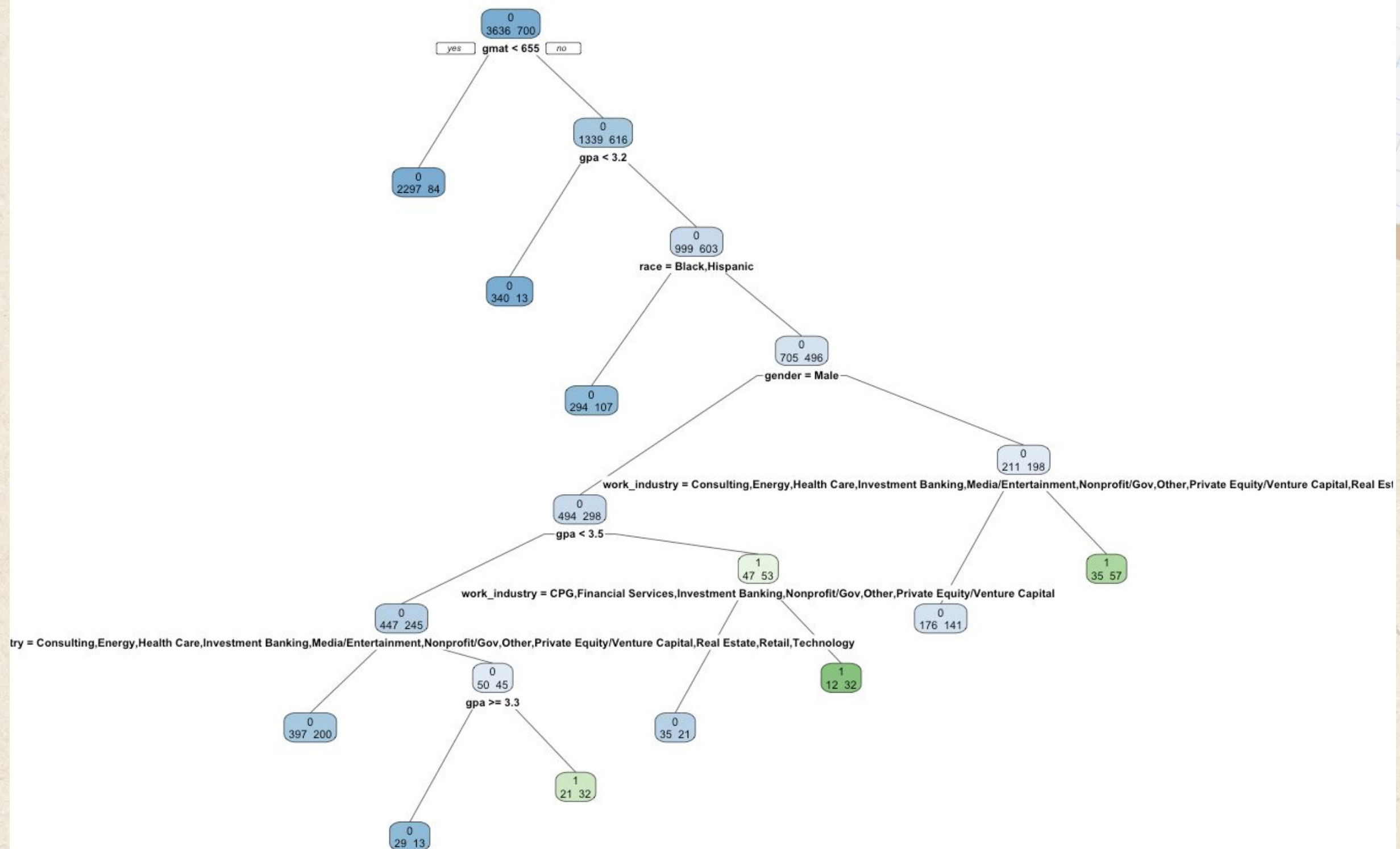
Significant predictors:

- (***) *genderMale, gmat*
- (*) *raceOther, work_industryInvestment Management, gpa:raceOther*

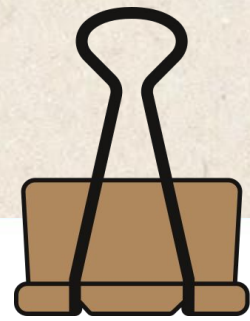


Decision Tree

84.18%
Accuracy



ADMITTED



Logistic Regression



88%
Accuracy

Training set: 70%

Validation set: 30%

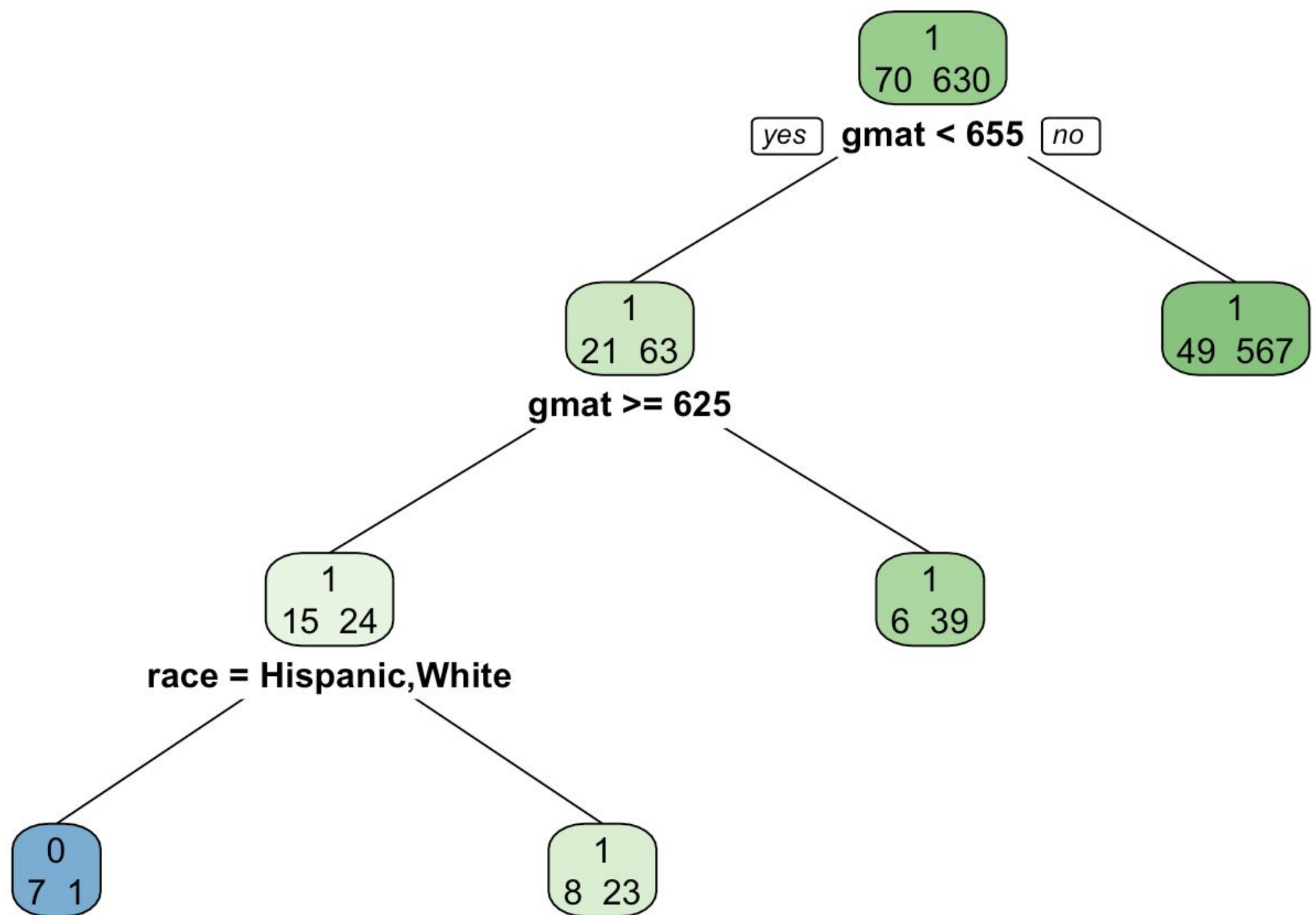
Predictors: *gender, international, gpa, major, race, gmat, work_exp, work_industry, gpa:race*

Significant predictors:

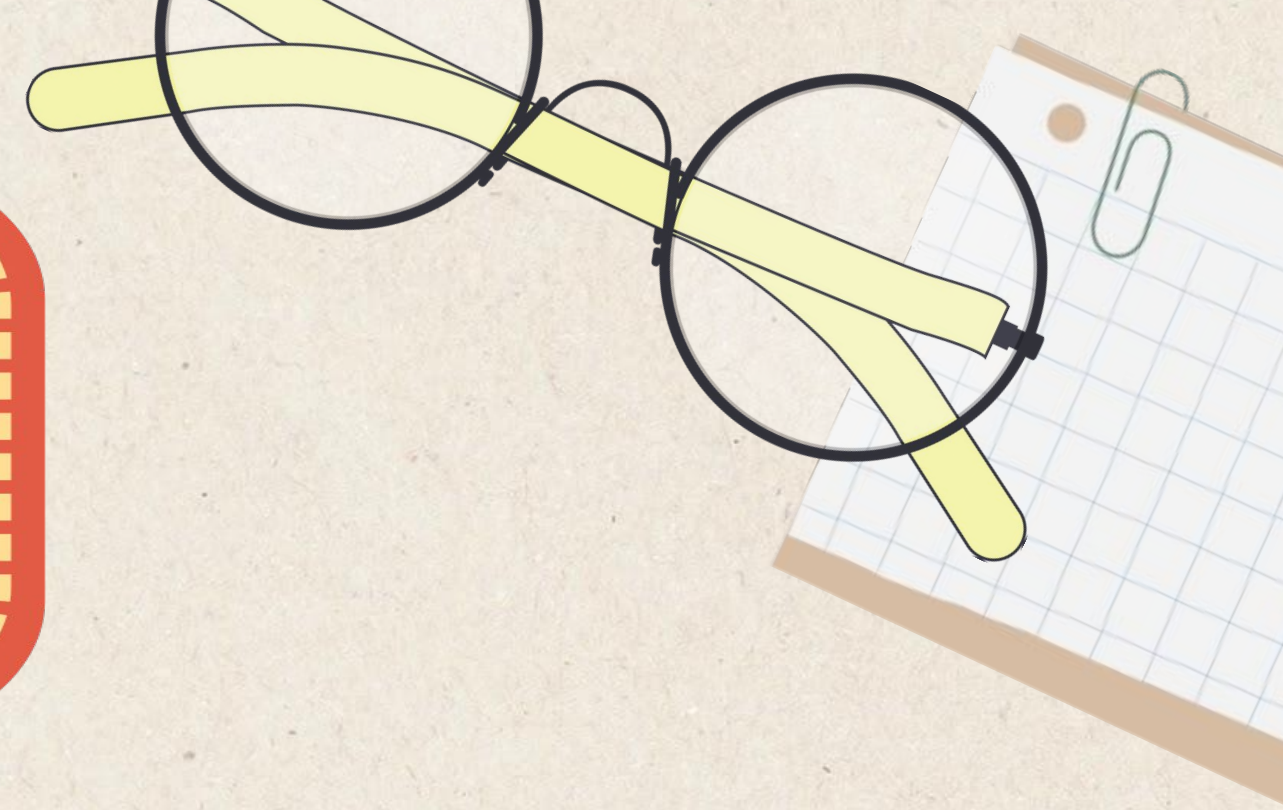
- (**) *gmat*
- (*) *genderMale, work_industryFinancial Services, work_industryTechnology*
- (.) *majorStem, work_industryInvestment Banking, work_industryInvestment Management, work_industryNonprofit/Gov*

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

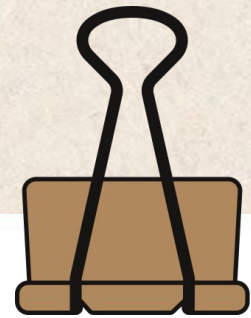
Rejected vs Not Rejected

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

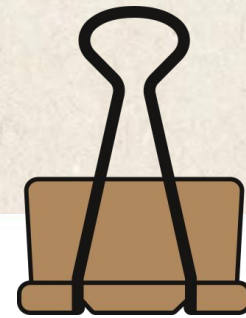
Admitted vs Waitlisted

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

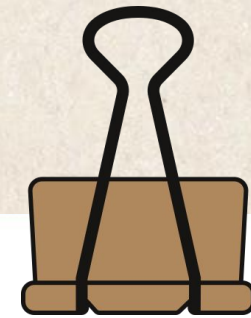
TAKEAWAYS



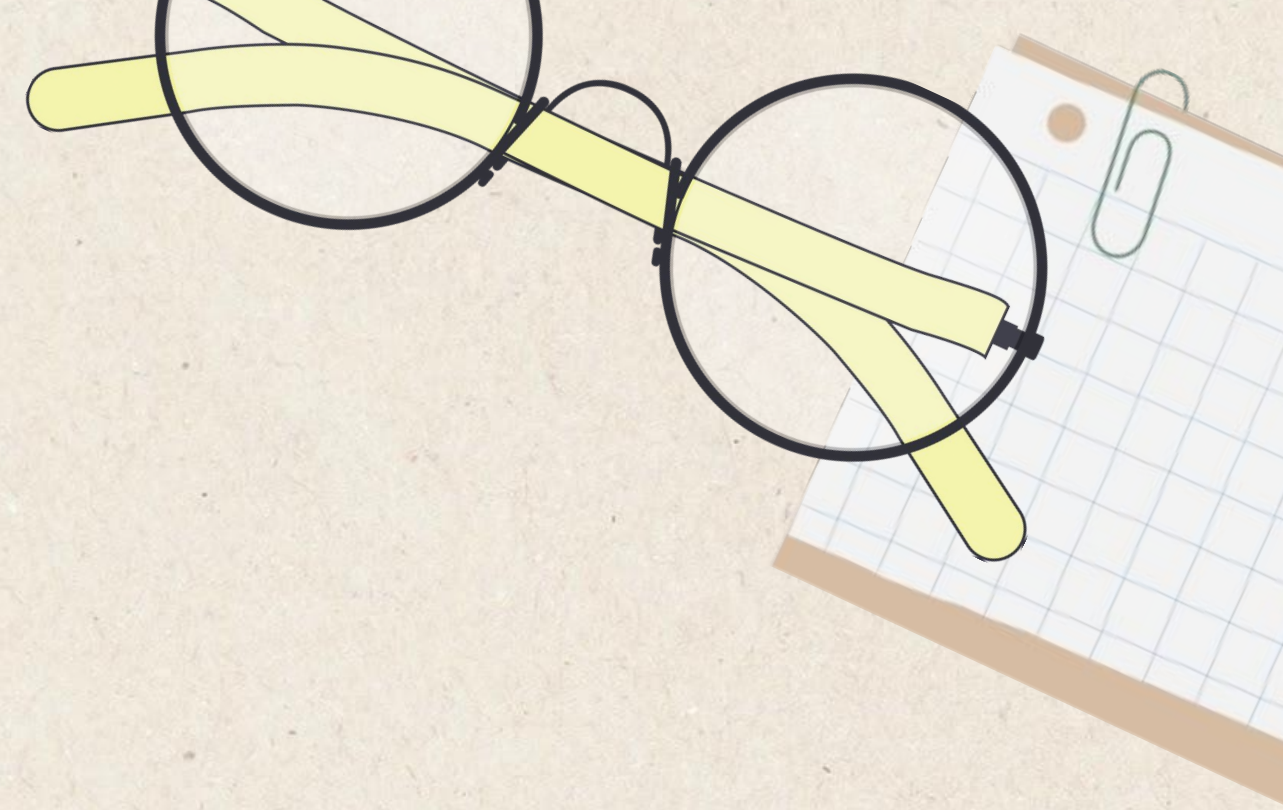
Relatively
accurate
models can be
created



Possible bias
for gender and
international
status



Could later
look more into
work industry
and race factors





Thank
you!

Any Questions?