# Data Mining Regressions

Michael Rose
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## Abstract

### Data

#### Overview of Data

The idea behind this dataset is to predict admissions into a Masters degree program. It was sampled from Engineering students at an Indian university. The parameters are the following:

parameter	range	description
GRE Score	0-340	Score on GRE exam
TOEFL Score	0 - 120	Score on TOEFL exam
University Ranking	0 / 5	Indian University Ranking
Statement of Purpose	0 / 5	Self assessed SOP score
Letter of Reccommendation	0 / 5	Self assessed LOR score
Undergraduate GPA	0 / 10	Cumulative undergraduate GPA
Research Experience	0 or 1	1 if Student engaged in research, 0 otherwise
Chance of Admit	$x \in [0, 1]$	Likelihood of admission

The source of this data is the following:

A Commonison of Domosocion Modele for Decidiotics of Conducts Admiration

A Comparison of Regression Models for Prediction of Graduate Admissions

Mohan S Acharya, Asfia Armaan, Aneeta S Antony

IEEE International Conference on Computational Intelligence in Data Science 2019

### Load Data

```
"TOEFL" = "TOEFL Score",

"Rating" = "University Rating",

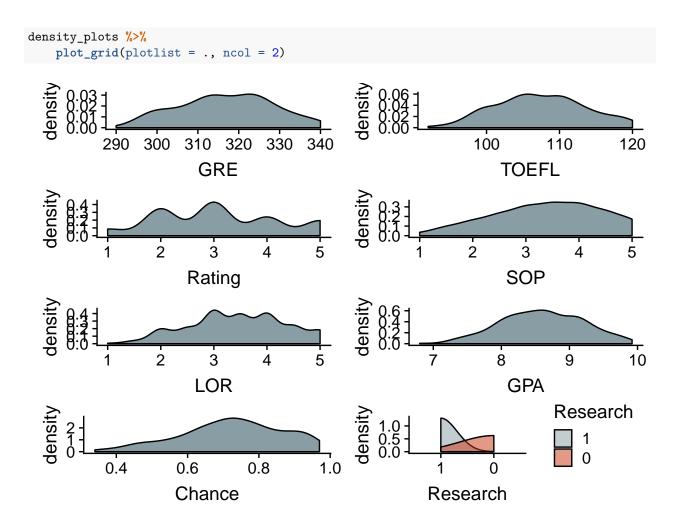
"GPA" = "CGPA",

"Chance" = "Chance of Admit") -> admissions
```

### Visualizations

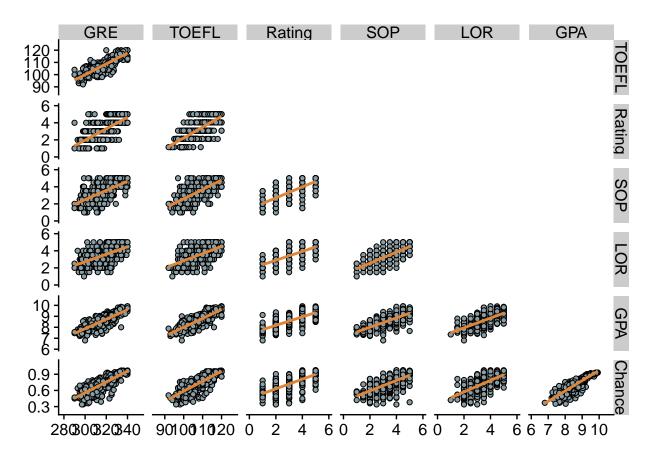
#### **Individual Features**

```
admissions
## # A tibble: 400 x 9
               GRE TOEFL Rating
                                 SOP
                                       LOR
##
     Student
                                             GPA Research Chance
##
       <int> <int> <int> <int> <dbl> <dbl> <dbl> <fct>
                                                           <dbl>
## 1
                             4
                                 4.5
                                                            0.92
          1
               337
                     118
                                       4.5 9.65 1
## 2
           2
                              4 4
                                       4.5 8.87 1
                                                            0.76
              324
                     107
## 3
           3 316
                     104
                              3
                                 3
                                       3.5 8
                                                            0.72
                                                 1
           4
## 4
             322
                    110
                              3 3.5
                                      2.5 8.67 1
                                                            0.8
## 5
           5 314
                             2 2
                                            8.21 0
                    103
                                       3
                                                            0.65
## 6
           6 330
                     115
                            5 4.5
                                       3
                                            9.34 1
                                                            0.9
           7
                             3 3
                                            8.2 1
## 7
               321
                     109
                                       4
                                                            0.75
## 8
           8
              308
                     101
                             2 3
                                       4
                                            7.9 0
                                                            0.68
## 9
           9
               302
                             1
                                 2
                     102
                                       1.5 8
                                                 0
                                                            0.5
## 10
          10
               323
                     108
                              3 3.5
                                      3
                                            8.6 0
                                                            0.45
## # ... with 390 more rows
# grab colnames
admissions %>% select(-c("Student", "Research")) %>% colnames() -> adm_colnames
adm_colnames
## [1] "GRE"
               "TOEFL" "Rating" "SOP"
                                         "LOR"
                                                  "GPA"
                                                           "Chance"
# make plotting function
plot_density <- function(variable){</pre>
 admissions %>%
   ggplot() +
     geom_density(aes(x = !!sym(variable)), fill = color_scheme[1])
}
# get density plots for each variable
density_plots <- future_map(adm_colnames, ~plot_density(.x))</pre>
# make a special density plot for binary Research variable
admissions %>%
 ggplot() +
 geom_density(aes(x = Research, fill = Research), alpha = 0.5) +
 scale_fill_manual(values = color_scheme) -> density_plots[[8]]
# plot
```



# **Combination of Predictors**

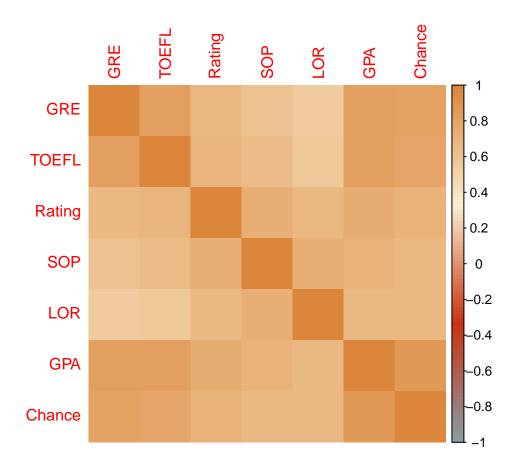
```
# make plotting function
plot_points <- function(data, mapping, ...){</pre>
  data %>%
    ggplot(mapping = mapping) +
    geom_point(fill = color_scheme[1], color = "black", pch = 21) +
    geom_smooth(method = "gam", color = color_scheme[4]) +
    scale_x_continuous(expand = expand_scale(mult = 0.3)) +
    scale_y_continuous(expand = expand_scale(mult = 0.3))
}
# grab lower plots from ggpairs
ggpairs_lower <- function(g){</pre>
  g$plots <- g$plots[-(1:g$nrow)]
  g$yAxisLabels <- g$yAxisLabels[-1]</pre>
  g$nrow <- g$nrow - 1
  g$plots <- g$plots[-(seq(g$ncol, length(g$plots), by = g$ncol))]
  g$xAxisLabels <- g$xAxisLabels[-g$ncol]</pre>
```



#### Correlations

```
# create color palette for corrplot
col_ramped <- colorRampPalette(color_scheme)

# select features to plot
admissions %>%
  select(-c("Student", "Research")) %>%
  cor() %>%
  corrplot(method = "shade", col = col_ramped(100))
```



We see that most of the predictor variables have relatively high correlation.

# Statistics

Algorithms

Exploration

Wrapup