# Predicting Texas Public Universities Retention Rate With Multi Variable Linear Regression and Neural Networks in R

Model's Training over 100 Epochs (Artificial Neural Network)

Percent\_of\_undergraduate\_students\_enrolled\_exclusivelv\_in\_distance\_education\_courses -7.155e-02 4.085e-02 -1.752 0.082476

Linear Model Output (Multi Variable Linear Regression)

Correlation of Full\_time\_retention\_rate

(Multi Variable Linear Regression)

6.294e-04 4.659e-04 1.351 0.179279

-8.703e-03 3.088e-03 -2.818 0.005673

-2.733e-02 3.214e-02 -0.850 0.396987

8.811e-03 4.704e-02 0.187 0.851751

4.513e-02 4.682e-02 0.964 0.337135

Correlation 0.4749923

0.2780218

0.7628461

0.5205848

-0.4205207 0.4921069

-0.3437749

0.5241315

-0.2783890

0.4502599

0.6709099

0.6744942 1.0000000

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# Problem Statement

## Research Questions:

What factors contribute most to student retention rates at Public Universities in Texas, and how can interventions be tailored to improve retention?

#### Significance and Relevance:

This research is important because it helps us figure out what helps students stay and succeed in Texas public universities. Increased retention will help both the universities and the students. Universities will generate more income and can use those profits to further develop the school and their programs. Additionally, students will be more likely to graduate with a high retention rate. With more income and more satisfied students the university will grow as a whole.

# Data Sources

#### Data Source

International Postsecondary Education Data System

#### **Data Content**

This full data set included all data imaginable for public universities. For 1 School over 1 year there is over 1,000 parameters or variables to look at. With such a large dataset I had to narrow down to factors that I believed may contribute to retention significantly either positively or negatively. With a larger scale project, I would consider pulling all the data to account for unseen factors.

### Descriptive Variables

- Institution\_Name
- Year

# Independent Variables

- Student\_to\_faculty\_ratio
- Full\_Time\_Staff\_per\_Student
- Average\_salary\_of\_full\_time\_professorsPercent\_of\_undergraduate\_enrollment\_Age\_18\_to\_24
- Percent\_of\_full\_time\_first\_time\_undergraduates\_ awarded\_any\_financial\_aid
- Total\_price\_for\_in\_state\_students\_living\_on\_campus
- Percent\_of\_undergraduate\_students\_enrolled\_ exclusively\_in\_distance\_education\_courses
- Percent\_admitted
- Published\_in\_state\_tuition\_and\_fees
- Books\_and\_supplies
- Undergraduate\_application\_fee
- SAT\_Math\_50th\_percentile\_score
- SAT\_Reading\_and\_Writing\_50th\_percentile\_score

#### Dependent

Full\_time\_retention\_rate

# Methods

#### Multi Variable Linear Regression

In my study, I preformed linear regression analysis to explore the relationship between the independent variables and university retention rates, dividing the dataset into a 70% training set and a 30% testing set for model development and validation. This methodology provided a systematic approach to identify and quantify the impact of key variables on retention rates, setting the stage for in-depth analysis and prediction.

#### **Artificial Neural Networks**

Additionally, I applied neural networks to predict university retention rates, using a processed and normalized dataset divided into training (70%) and testing (30%) sets. This approach enabled me to explore complex patterns and relationships between factors affecting retention rates. By training and evaluating the neural network model, I gained insights into its predictive power and the impact of specific variables on student retention.

Min 1Q Median 3Q Max -14.1680 -2.8956 0.8128 3.9232 12.2735

otal\_price\_for\_in\_state\_students\_living\_on\_campus

AT\_Reading\_and\_Writing\_50th\_percentile\_score

Published\_in\_state\_tuition\_and\_fees

ndergraduate\_application\_fee

T\_Math\_50th\_percentile\_score

tudent\_to\_faculty\_ratio

Percent\_admitted

Books\_and\_supplies

Full\_time\_retention\_rate

Scatter Plot of Independent Variables

vs Retention Rate (MVLR)

Full\_Time\_Staff\_per\_Student

Average\_salary\_of\_full\_time\_professors

Published\_in\_state\_tuition\_and\_fees

Undergraduate\_application\_fee

SAT\_Math\_50th\_percentile\_score

Percent\_of\_undergraduate\_enrollment\_Age\_18\_to\_24

Total\_price\_for\_in\_state\_students\_living\_on\_campus

SAT\_Reading\_and\_Writing\_50th\_percentile\_score

Percent\_of\_full\_time\_first\_time\_undergraduates\_awarded\_any\_financial\_aid

Percent\_of\_undergraduate\_students\_enrolled\_exclusively\_in\_distance\_education\_cour

Books\_and\_supplies

# Results

#### Multi Variable Linear Regression

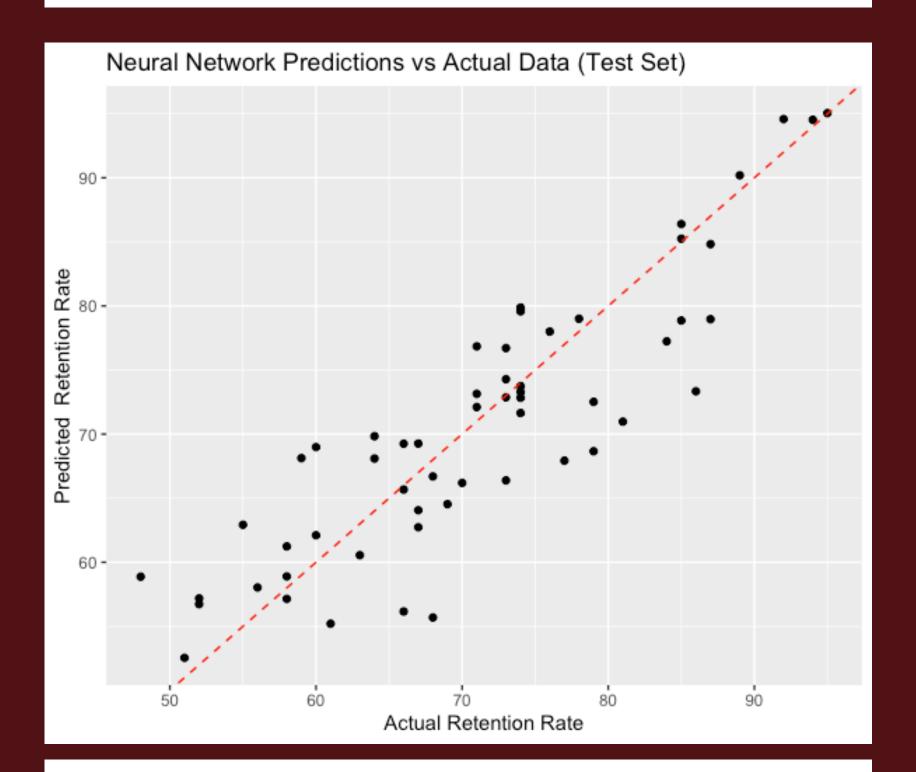
I found that the average salary of full-time professors was the most significant factor affecting full-time retention rates. This further highlighted the strong correlation between faculty compensation and student retention found in the correlation matrix. Additionally, higher SAT Reading and Writing, as well as Math 50th percentile scores, were significantly associated with improved retention rates, suggesting the importance of academic preparedness.

#### Artificial Neural Networks

In my project, the neural network model exhibited strong predictive performance with a Mean Absolute Error (MAE) of 0.414 and an adjusted R-squared value of 0.768, effectively capturing the variance in full-time retention rates. This accuracy, highlighted by the low Mean Squared Error (MSE) of 0.300 and Root Mean Squared Error (RMSE) of 0.547, shows the model's robustness in utilizing selected factors to predict outcomes.

# Multi Linear Regression Predictions vs Actual Data (Test Set)

Testing Model's Statistical Outputs (Artificial Neural Network)



Testing Model's Statistical Outputs (Artificial Neural Network)

# Findings

The findings from my research reveal critical insights into improving student success. Through linear regression analysis, I identified the average salary of full-time professors, SAT Reading and Writing, Math 50th percentile scores, and the percentage of undergraduates aged 18 to 24 as significant predictors of full-time retention rates, underscoring the importance of faculty compensation and academic preparedness. The application of an artificial neural network further validated these findings, demonstrating strong predictive performance with an R-squared value of 0.7872, thus highlighting the model's capability to accurately predict retention rates based on selected variables. This comprehensive analysis suggests that strategic interventions focusing on enhancing faculty salaries, student academic readiness, and catering to the needs of traditional-aged undergraduates could significantly bolster retention at Texas public universities.

# Implications

This study's results underscore the importance of faculty compensation, academic preparedness, and demographic considerations in student retention at Texas public universities, providing actionable insights for targeted interventions. Enhancing professor salaries, bolstering student readiness, and focusing on the needs of traditional-aged students can lead to significant improvements in retention rates. These findings offer a direct pathway to addressing the research questions posed, with significant implications for policy and practice aimed at boosting student success and institutional development.

# References/Data Cite

International Postsecondary Education Data System Overall Website: <a href="https://nces.ed.gov/ipeds/">https://nces.ed.gov/ipeds/</a>
Data Source:

https://nces.ed.gov/ipeds/datacenter/InstitutionByName. aspx?goToReportId=5&sid=6d91ee2e-4f45-4b44bca7-7fb33c561d87&rtid=5

