



NBA Salary Prediction & Analysis

i Information available in audio.

Deep diving into our data with clear direction

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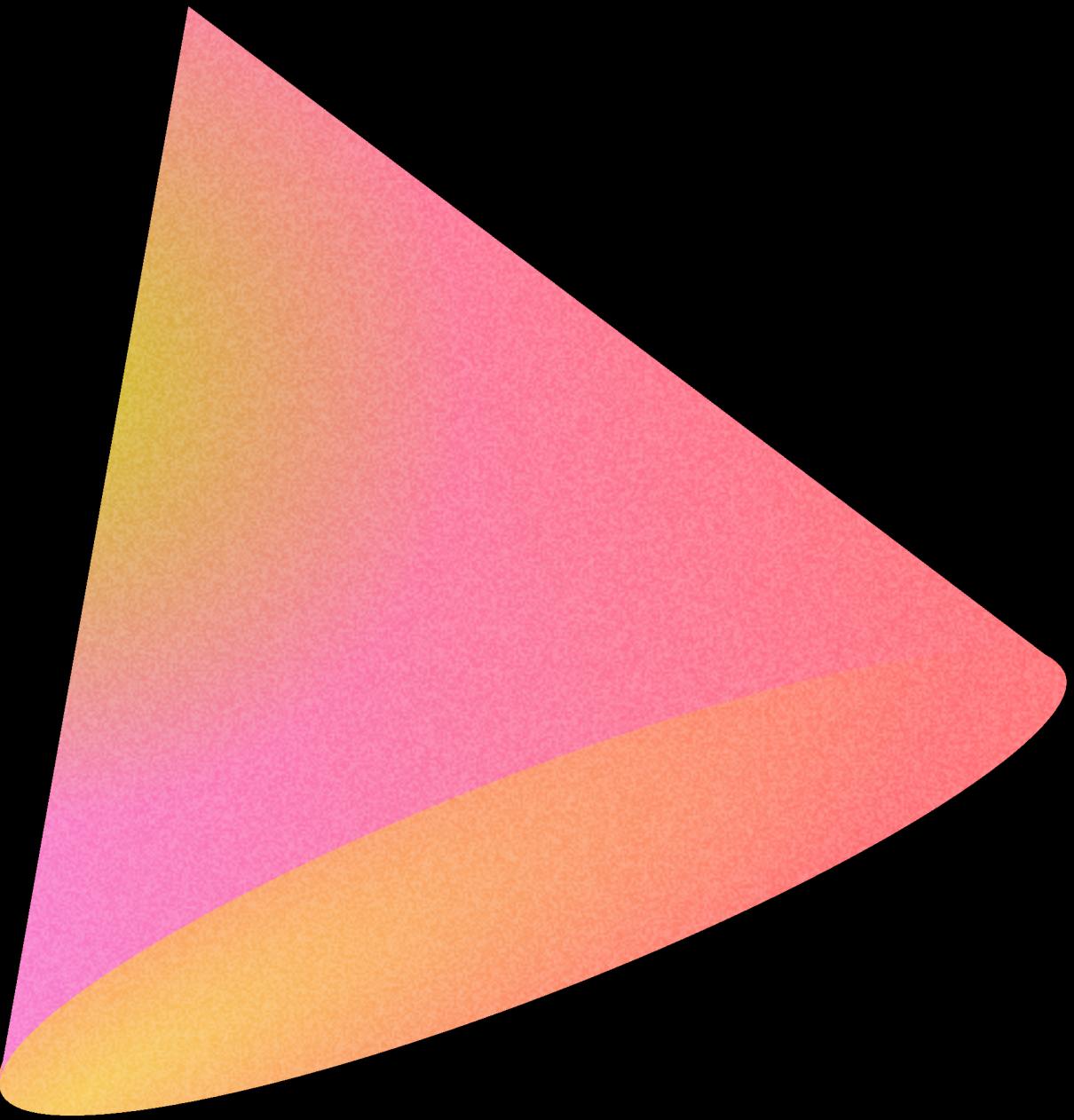
Introduction

Purpose:

To predict NBA salaries for college basketball players based on their performance statistics and historical data, aiding NBA teams in draft and salary decisions.

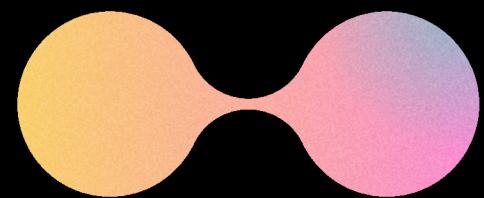
Description:

This project tackles the challenge of translating college performance into professional success, using machine learning and comprehensive data from current and former NBA players to improve player evaluation and financial planning.



Data Resources & Collection

- College Basketball Player's Performance Data
- Historical Performance Data
- NBA Draft Combine Data
- Salary Data



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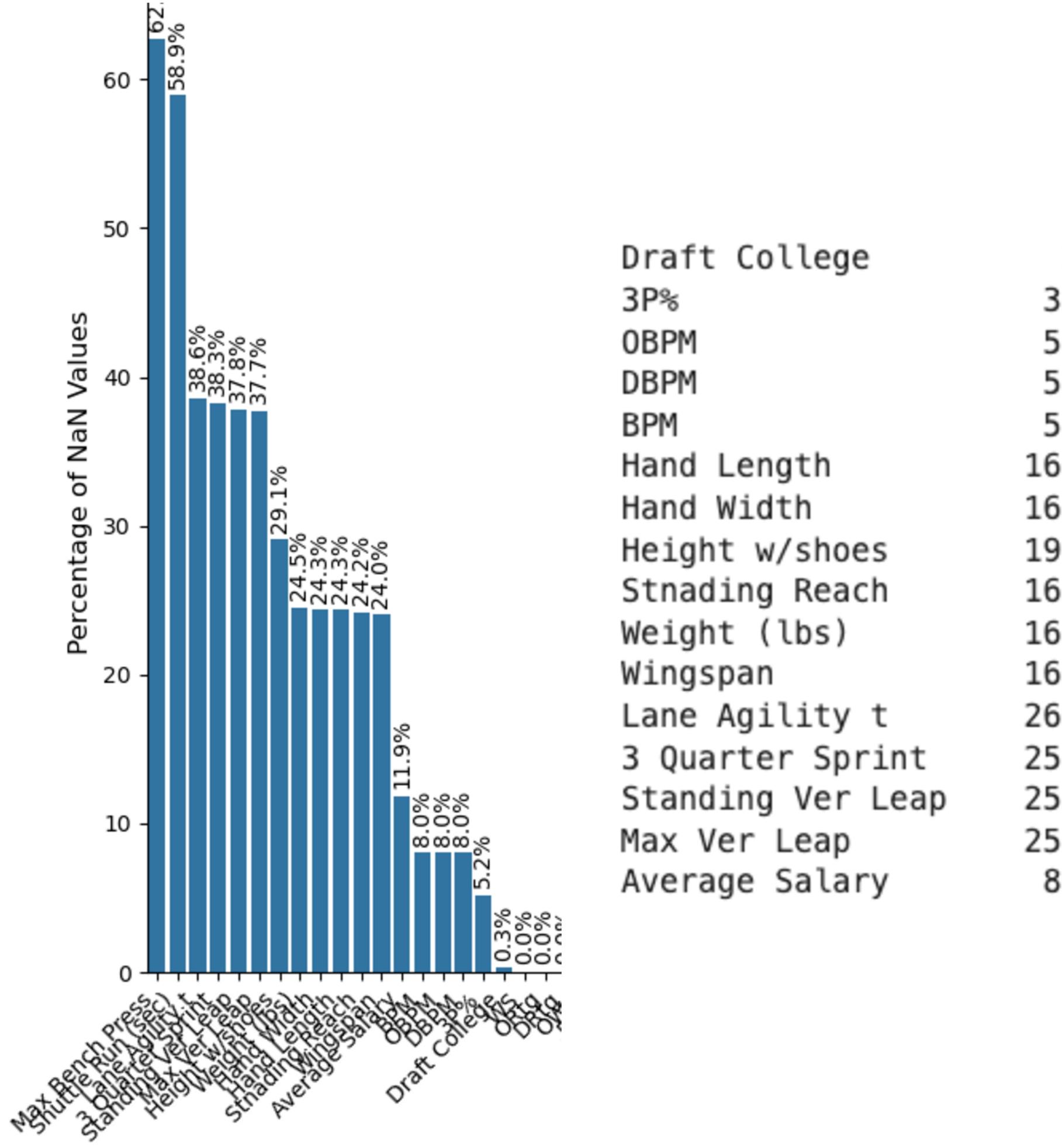
Process:

- Data time frame 2009-2024
- College Basketball Performance Data was obtained from <https://www.sports-reference.com/>
- NBA Draft Combined was collected using NBA Api from <https://www.nba.com/stats/draft/combine>
- Salary Data was obtained from HoopsHype using Selenium web scraper
<https://hoopshype.com/salaries/>

Issues During Collection & Cleaning:

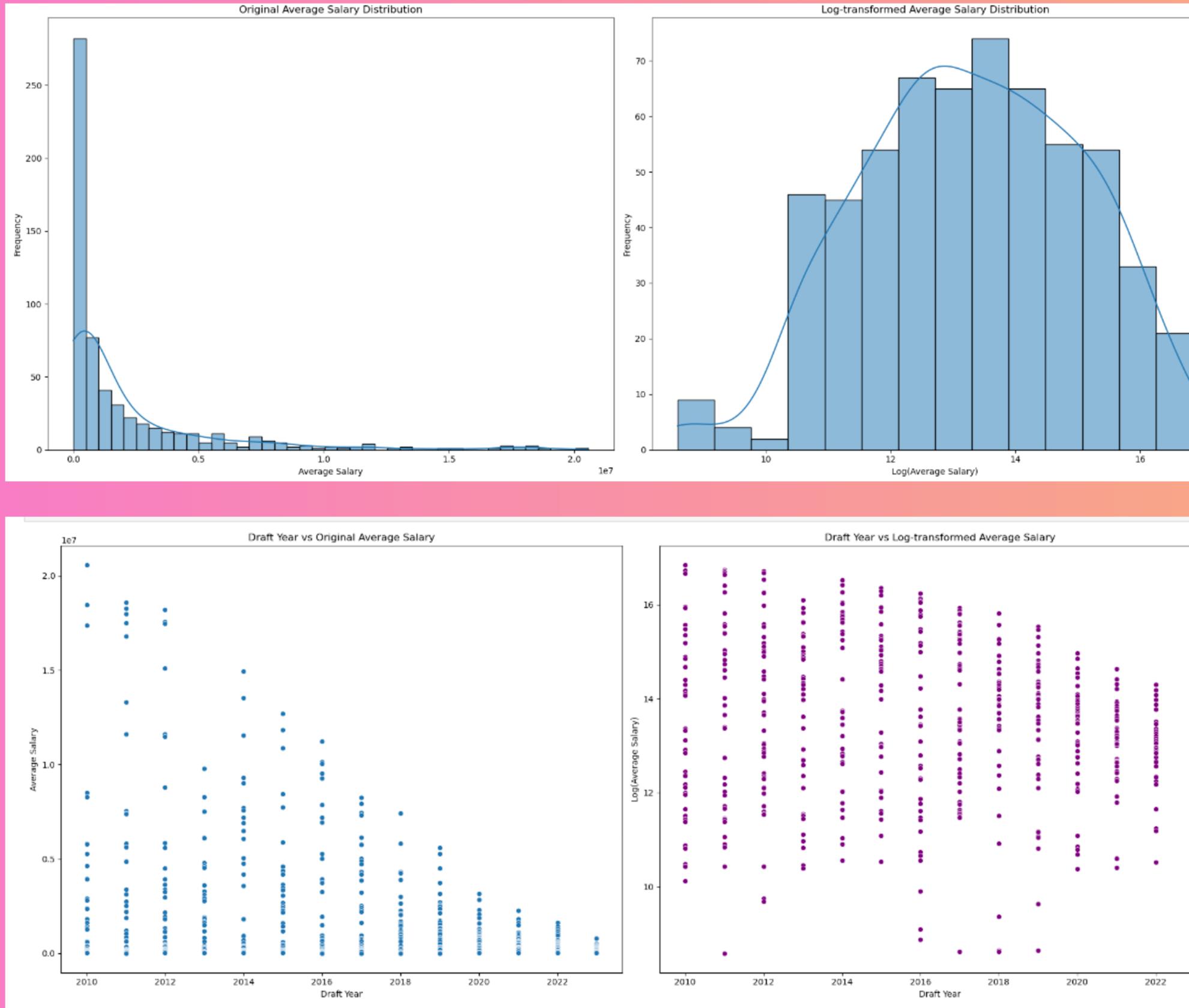
- Six different csv files to combine
- All data is in different format. Data types had to be changed
- NBA Draft Combine data missing a lot of values.
- Salary Data has missing values due to the issue with web scraper

Missing Values: Drop or Impute



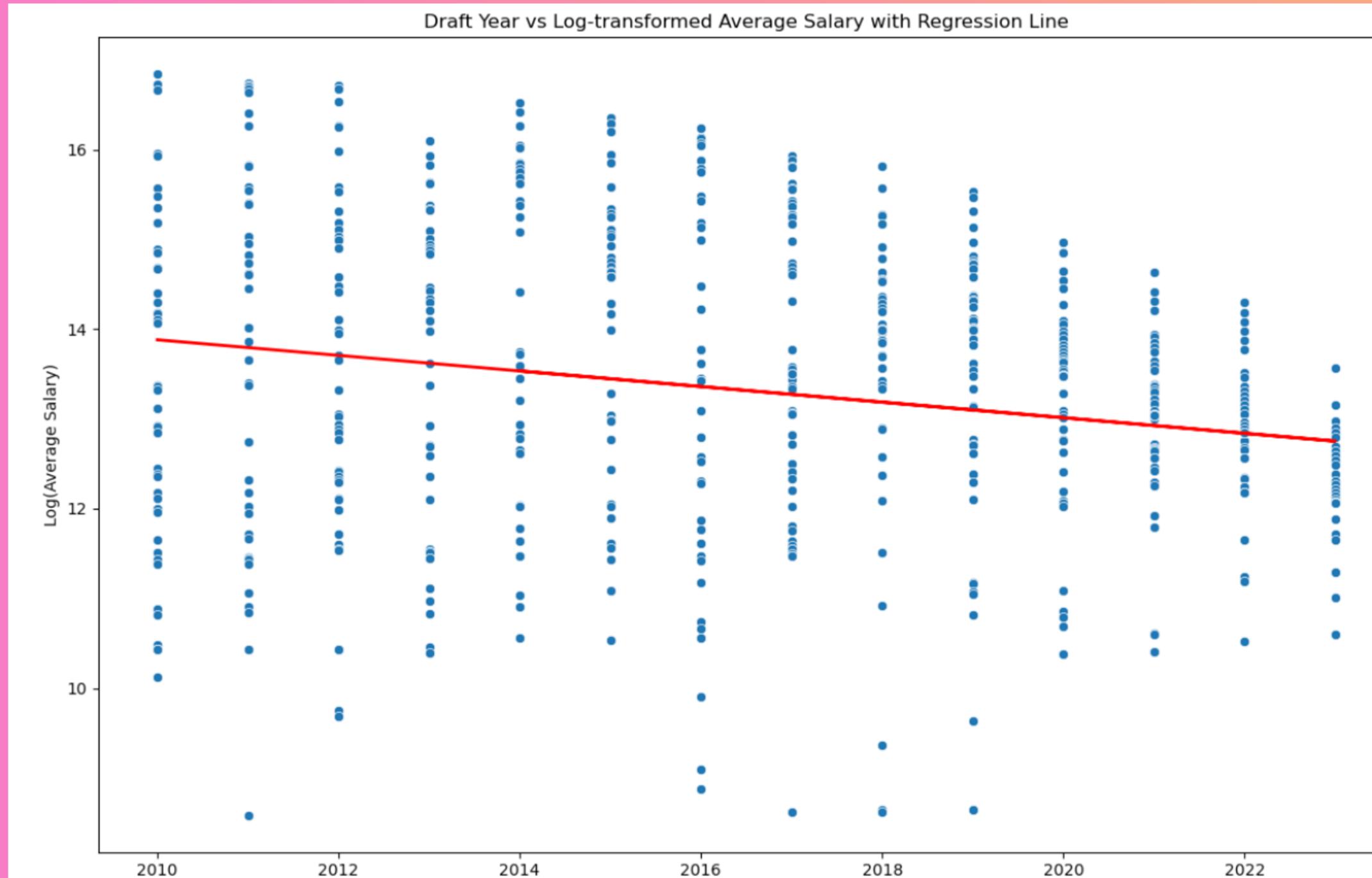
- Very high percentage of values for Max Bench Press and Shuttle Run
- Draft College fixed by adding data
- NBA Draft Combine
- Average Salary

Average Salary: Imputation



- Before Log Transformation
 - Skewed to the right
 - Skewness 2.82
 - High salary outlier
- After Log Transformation
 - Data more normally distributed
 - Skewness -0.17
 - The difference between salaries is more proportional

Average Salary: Linear Regression Draft Year and Salary



- **Observations and Conclusions:**
 - Pearson correlation coefficient is **-0.2106**
 - P-value is small **0.0000 < 0.0001**
 - The weak correlation suggests that we can not just use Draft Year alone to predict players salaries.
 - Other factors should be also looked at

Next Steps:

- Finish cleaning my dataset. Impute other missing values. Hot-encode categorical variables.
- Feature Selection/Engineering:
 - Identify key college performance metrics that might influence NBA salary (e.g., points per game, rebounds, assists, shooting percentages).
- Create new features if needed (e.g., composite scores, efficiency ratings).
- Model selection: Multiple Regression, Ridge Regression, Random Forest, Neural Networks.



Questions?



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