R vs. Python: Data Science Comparison

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Presentation Overview

- SYE Goals
- Olympics Project Overview + Findings
- NBA All-Star Project Overview + Findings
- Conclusion

SYE Goals

- Learn Python for data science
- Go through a Python data science textbook and explore Positron IDE
- Compare R and Python functions side by side
- Be able to identify strengths of both languages, and combine them in a project

What is R? What is Python?



Positron



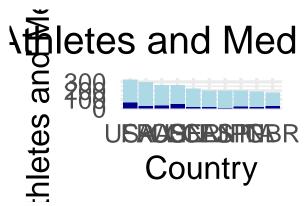
- An IDE that is in development
- Allows easy usage of both R and Python

Olympics Project Overview

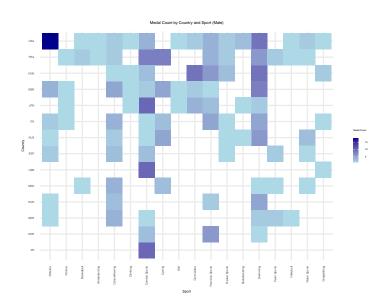
- Split the semester into two main projects: 2024 Paris Olympics and NBA
- First: Wrangle data and generate plots in R
- Next: replicate everything in Python, following along with a textbook called "Python for Data Science"

Some Olympics Project Plots

- Light blue represents number of athletes each country sent to the Olympics
- Dark blue represents the number of medals won



Some Olympics Project Plots



R vs. Python Code Comparison

 Here is how I performed the joining of the male athletes and male medals datasets before creating the bar plot using R:

• And here is how I did that in Python:

```
Male_Athletes_Count_joined = pd.merge(
  Male_Athletes_Count, Male_Total_Medal_Counts,
  on="country_code", how="left")
```

R vs. Python Code Comparison

- Before making the male bar plot, I had to group by country and add up all the athletes for each country.
- Here is how I did it in R:

```
Male_Athletes_Count <- Male_Athletes_df |>
  group_by(country_code) |>
  summarise(athlete_count = n())
```

• And here is how I did it in Python:

```
Male_Athletes_Count = Male_Athletes_df.groupby(
"country_code").size().reset_index(name="athlete_count")
```

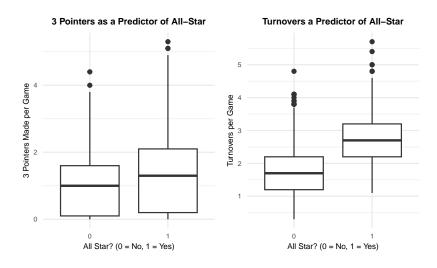
NBA Project

- The second project I worked on was predicting whether NBA players made the all-star game based on their stats
- The dataset included all player statistics each season from 1947 to present
- First part of the project: wrangle and plot in R
- Second part: perform a statistical learning method (KNN) in Python

NBA Project Results

- Interestingly, when making the exploratory plots, one of the variables that looked the most correlated with players making the all-star game was having a high average turnovers
- The variables I used in the KNN model were: points, rebounds, assists, steals, field goals attempted, and turnovers
- Final KNN model accuracy was 88.22%

Turnovers and Three Pointers as Predictors of All-Star



Confusion Matrix

- Confusion matrix with the results of my KNN model
- Top left number, 559, gives us the number of non-all-stars that our knn model correctly predicted as non-all-stars
- Bottom right number → number of all-stars that our model correctly predicted as all-stars.
- \bullet Bottom left \to non-all-stars that our model predicted to actually be all-stars
- Top right → players that our model predicted to not be all-stars but were actually all-stars

```
# Create and display the confusion matrix
conf_matrix = sklearn.metrics.confusion_matrix(
  test_cat, knn_mod)
print(conf_matrix)

## [[559 14]
## [ 70 70]]
```

Thoughts on Learning Python for Data Science

- I am surprised that Python is used so much for data science outside of SLU
- R seems more intuitive and organized, especially when it comes to data wrangling
- As an introduction to coding or a machine learning class, I believe Python makes more sense to learn, but I'm glad our department teaches R for DATA234

Conclusion

- In my opinion, R is easier to learn for data science, especially for beginner coders
- Both languages have similar math/statistical capabilities
- For data wranging and plotting, R is great!
- Python may be better (or at least equal) in terms of stat.
 learning