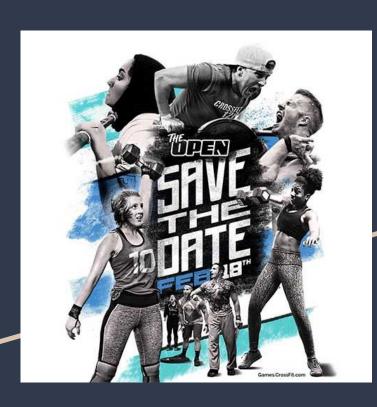
2020 Crossfit Open Analysis



Scott Graham 24th June, 2022

Data Summary



The Crossfit Open is the largest single sporting competition held worldwide. The goal, to find the fittest athletes on Earth.

Using the athlete metrics and workout results for 230,000+ for the five workouts we hope to be able to determine some key features of top performing athletes

Outline

- Purpose of Research and Hypothesis
- Data Understanding
- Data Cleaning
- Regression Analysis
- Further Analysis
- Conclusion

Purpose of Research and Hypothesis

To determine if athlete metrics, including age, height and weight are key determinants of better results

Difference between male and female body metrics

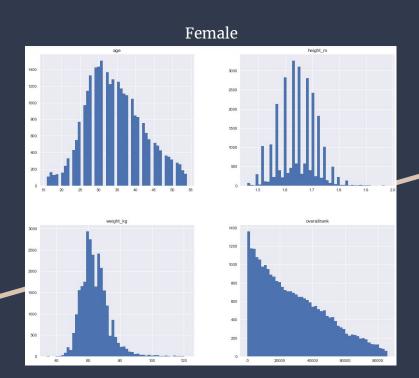
Comparison of the very top athletes vs the general population and difference in body metrics

Data Understanding

Two datasets were used and imported from Kaggle:

- 1. 2020 Athletes (393,000+ rows, 19 columns)
 Provided all the general information about the athlete including:
 - a. Name
 - b. Gender
 - c. Age
 - d. Weight
 - e. Height
 - f. Division they compete in
 - g. Overall rank and score
- 2. 2020 Workouts (1965000+ rows, 13 columns) Provided information regarding:
 - a. Gym they did the workouts in
 - b. Their judge
 - c. if the workout was scaled or as prescribed
 - d. Results and ranking of each workout

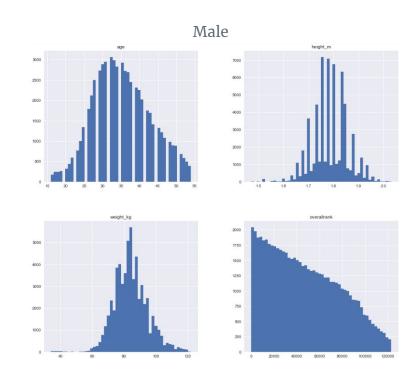
Data Cleaning



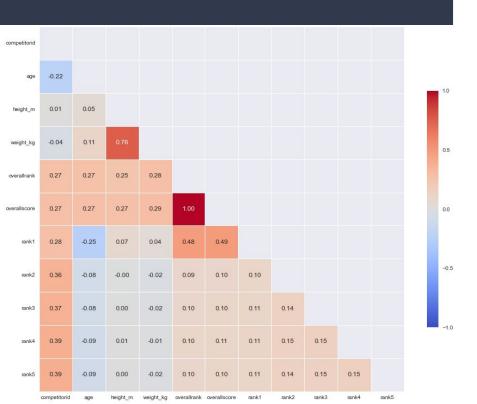
Removal of columns with no relevance to goal and duplicate rows

I removed all athletes that did not have both height and weights accurately entered (eg. 1kg body weight)

Seperating male and female athletes



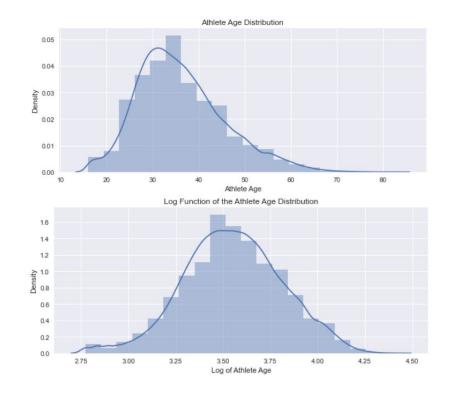
Data Cleaning



Removal of outliers using Z-Score

Normalise age data for easier use

Compare data against each other using a correlation matrix



Regression Analysis

Train-Test-Split done with 70/30 split

Train and Test Mean Square Error converged:

- Train Mean Squarred Error: 0.8786019
- Test Mean Squarred Error: 0.8900844
- Meaning that our data was producing similar results in the training model to the test model

However our R-Squared value was only 0.155, meaning our data had only a 15.5% chance of predicting the athletes overall rank from their age, weight, height and gender

Further Analysis

Female Top Quartile age 32.39 height_m 1.64 weight_kg 62.95 overallrank 4,837.77 dtype: float64

Female Second Quartile
age 34.17
height_m 1.65
weight_kg 63.09
overallrank 16,763.24
dtype: float64

Female Third Quartile
age 35.43
height_m 1.65
weight_kg 63.76
overallrank 32,737.57
dtype: float64

Female Bottom Quartile age 35.84 height_m 1.65 weight_kg 66.27 overallrank 57,667.13 dtype: float64 Games Female Athletes age 29.35 height_m 1.64 weight_kg 64.21 overallrank 2,933.21 dtype: float64 Split data into the 4 quartiles based on overall ranking of athletes

Take the mean values of age, height and weight of each quartiles and compare to Games athletes

Male Top Quartile
age 31.86
height_m 1.78
weight_kg 83.64
overallrank 9,554.44
dtype: float64

Male Second Quartile
age 34.63
height_m 1.78
weight_kg 83.02
overallrank 30,798.78
dtype: float64

Male Third Quartile
age 36.74
height_m 1.79
weight_kg 84.20
overallrank 56,091.27
dtype: float64

Male Bottom Quartile age 38.25 height_m 1.79 weight_kg 86.17 overallrank 90,082.63 dtype: float64 Games Male Athletes
age 28.99
height_m 1.77
weight_kg 86.43
overallrank 4,010.39
dtype: float64

Conclusions



Unable to produce a predictive model based on athlete metrics, this is obviously due to many other parameters around fitness performance

The performance of athletes could be seen to decrease as they aged. With Games athletes being the youngest on average

Future Work

Better use of predictive modelling with expansion on workout information would be a great way to advance this analysis

Thank You!

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