# Project Proposal <br> Sport Stat - Olympics Data 

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As a scientist in the Sport field, we are asked to do deep analysis which will be focused through data visualization
The analysis will find out any trends in Olympics games such as the country that dominating in certain sports for 120 years and qualitative analysis to answer the question why this phenomenon happens

Further, our journey steps to the side of statistics with regression problem to estimate the missing value in certain variables

It's interesting to find out specific, rightful, and useful regression method to handle this problem

## ASSUMPTIONS

The scopes of research are listed:

## FIRST

Each rows is the unique person in difference time of period

## SECOND

The chosen columns must have high correlation

## QUESTIONS

To ensure the effectiveness of the research, the following question will be answered systematically:


## APPROACH

To ensure the effectiveness of the research, the following question will be answered systematically:

Several columns will be dropped because it has no enough correlation with the main analysis

To handle missing value, the linear regression and decision tree regressor will be compared

Evaluation metrics use root mean square error, mean absolute error, and Pearson correlation

Explanatory Data Analysis (EDA) is rightful method and mostly used to find out pattern in the whole data


Two tables are included. But for the main analysis, we only need the athlete event table

Height
Weight

## EXPLORATION

|  | ID | Name | Sex | Age | Height | Weight | Team | NOC | Games | Year | Season | City | Sport | Event | Medal |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 1 | A Dijiang | M | 24.0 | 180.0 | 80.0 | China | CHN | $1992$ <br> Summer | 1992 | Summer | Barcelona | Basketball | Basketball Men's Basketball | NaN |
| 1 | 2 | A Lamusi | M | 23.0 | 170.0 | 60.0 | China | CHN | $2012$ <br> Summer | 2012 | Summer | London | Judo | Judo Men's ExtraLightweight | NaN |
| 2 | 3 | Gunnar Nielsen Aaby | M | 24.0 | NaN | NaN | Denmark | DEN | $1920$ <br> Summer | 1920 | Summer | Antwerpen | Football | Football Men's Football | NaN |
| 3 | 4 | Edgar Lindenau Aabye | M | 34.0 | NaN | NaN | Denmark/Sweden | DEN | $1900$ <br> Summer | 1900 | Summer | Paris | Tug-OfWar | Tug-Of-War Men's Tug-Of-War | Gold |
| 4 | 5 | Christine Jacoba Aaftink | F | 21.0 | 185.0 | 82.0 | Netherlands | NED | $\begin{array}{r} 1988 \\ \text { Winter } \end{array}$ | 1988 | Winter | Calgary | Speed Skating | Speed Skating Women's 500 metres | NaN |


| $\begin{aligned} & 0 \\ & 0 \end{aligned}$ | <class 'pandas.core.frame. DataFrame'> |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | RangeIndex: 271116 entries, 0 to 271115 |  |  |  |  |
|  | Data | columns | (total | 15 column |  |
|  | \# | Column | Non-Nul | 1 Count | Dtype |
| ㄷ | 0 | ID | 271116 | non-null | int64 |
|  | 1 | Name | 271116 | non-null | object |
| Ј | 2 | Sex | 271116 | non-null | object |
| $\bigcirc$ | 3 | Age | 261642 | non-null | float64 |
| U | 4 | Height | 210945 | non-null | float64 |
| - | 5 | Weight | 208241 | non-null | float64 |
| 0 | 6 | Team | 271116 | non-null | object |
| ᄃ | 7 | NOC | 271116 | non-null | object |
| O | 8 | Games | 271116 | non-null | object |
| ¢ | 9 | Year | 271116 | non-null | int64 |
| 으 | 10 | Season | 271116 | non-null | object |
| 는 | 11 | City | 271116 | non-null | object |
| U | 12 | Sport | 271116 | non-null | object |
| (1) | 13 | Event | 271116 | non-null | object |
| - | 14 | Medal | 39783 n | non-null | object |
|  | dtypes: float64(3), int64(2), object(10) |  |  |  |  |


| ID | 0 |
| :---: | :---: |
| Name | 0 |
| Sex | 0 |
| Age | 9474 |
| Height | 60171 |
| Weight | 62875 |
| Team | 0 |
| NOC | 0 |
| Games | 0 |
| Year | 0 |
| Season | 0 |
| City | 0 |
| Sport | 0 |
| Event | 0 |
| Medal dtype: | $\begin{aligned} & 231333 \\ & t 64 \end{aligned}$ |


| $\xrightarrow{\sim}$ | ID | 135571 |
| :---: | :---: | :---: |
| E | Name | 134732 |
| Ј | Sex | 2 |
| $\bigcirc$ | Age | 74 |
| $\bigcirc$ | Height | 95 |
| ᄃ | Weight | 220 |
| $\bigcirc$ | Team | 1184 |
| (1) | NOC | 230 |
| (1) | Games | 51 |
| T | Year | 35 |
| $>$ | Season | 2 |
| 1 | City | 42 |
| $\bigcirc$ | Sport | 66 |
| C | Event | 765 |
| $\bigcirc$ | Medal dtype: | $64^{3}$ |

## Findings

Three important variables for deep analysis need manipulation. These are athlete's age, weight, and height. So, it needs to find out the best method to fill those missing value properly

## EXPLORATION





## Findings

According to those histogram, athlete's age and weight are right-skewed while the height is bell-shape, Normal distribution.

- The average of athlete's age is about 97 yo. It's unnatural. So, we need to do pre-processing
- The maximum of athlete's weight is about 214 kg . This is why the histogram would be rightskewed


## EXPLORATION





## Findings

The correlation is the statistic indicating the relationship between two variables in the data. After exploring the numerical variables, the correlation between athlete's weight and height is high.


| ᄃ RMSE in CV - 1: 8.774727 and MAE: 6.258586 |  |
| :---: | :---: |
|  | RMSE in CV - 2: 8.55525 and MAE: 6.065735 |
| $\stackrel{\square}{\square}$ | RMSE in CV - 3: 8.793303 and MAE: 6.19477 |
| 으 | RMSE in CV - 4: 8.461461 and MAE: 6.039466 |
| రె | RMSE in CV - 5: 8.589853 and MAE: 6.084789 |
|  | RMSE in CV - 6: 8.575798 and MAE: 6.123329 |
|  | RMSE in CV - 7: 8.577335 and MAE: 6.086527 |
| 0 | RMSE in CV - 8: 8.65246 and MAE: 6.145842 |
| U | RMSE in CV - 9: 8.91851 and MAE: 6.243718 |
|  | RMSE in CV - 10: 8.766976 and MAE: 6.186461 |
| $\bigcirc$ | Average of RMSE: 8.666567278592114 |
|  | Average of MAE: 6.142922312460948 |


|  | Age | Height | Weight |
| ---: | ---: | ---: | ---: |
| count | 261642.000000 | 210945.000000 | 208241.000000 |
| mean | 25.556898 | 175.338970 | 70.702393 |
| std | 6.393561 | 10.518462 | 14.348020 |
| min | 10.000000 | 127.000000 | 25.000000 |
| $\mathbf{2 5 \%}$ | 21.000000 | 168.000000 | 60.000000 |
| $\mathbf{5 0 \%}$ | 24.000000 | 175.000000 | 70.000000 |
| $\mathbf{7 5 \%}$ | 28.000000 | 183.000000 | 79.000000 |
| $\mathbf{m a x}$ | 97.000000 | 226.000000 | 214.000000 |

## Findings

The RMSE of prediction is about 8.66 where it is comparable with the standard deviation of response variable. So, the linear regression model is quite good. The model equation is:

Intercept: -118.8526691879782<br>Coefficient: 1.08081366279174




|  | fit_time | std_fit_time | mean_score_time | std_score_time | param_max_depth | param_min_samples_leaf | param_min_samples_split | params |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 0.133030 | 0.027070 | 0.004408 | 0.000763 | 10 | 2 | 2 | \{max_depth': 10, 'min_samples leaf: 2, 'min. |
| 1 | 0.156390 | 0.027086 | 0.005046 | 0.001747 | 10 | 2 | 50 | [max depth': 10 , 'min_samples leaf: 2, 'min |
| 2 | 0.186011 | 0.035946 | 0.005567 | 0.001250 | 10 | 2 | 75 | \{max_depth': 10 , 'min_samples leaf: 2, 'min_. |
| 3 | 0.150480 | 0.019991 | 0.004860 | 0.001055 | 10 | 2 | 100 | \{max_depth': 10, 'min_samples_leaf: 2, 'min_. |
| 4 | 0.162655 | 0.043891 | 0.004723 | 0.000673 | 10 | 2 | 120 | \{'max_depth': 10, 'min_samples_leaf: 2, 'min |

## Grid-search to get optimum hyper parameters

```
Best hyperparameters :
    {'max depth': 10, 'min_samples_leaf': 100, 'min samples split': 2}
Best evaluation :
    -8.634977365979429
Best model of Decision Tree:
    DecisionTreeRegressor(max_depth=10, min_samples_leaf=100)

\section*{PRE-PROCESSING}

Regression Model RMSE Training RMSE Validation MAE Training MAE Validation Pearson Training Pearson Validation
\begin{tabular}{rrrrrrr}
\hline Linear Regression & 8.66746 & 8.68068 & 6.14282 & 6.1264 & 0.79574 & 0.79806 \\
Decision Tree Baseline & 8.62923 & 8.64788 & 6.11897 & 6.11031 & 0.79777 & 0.79977 \\
Decision Tree Grid-Search & 8.63171 & 8.65194 & 6.12056 & 6.11259 & 0.79764 & 0.79956
\end{tabular}

Linear regression is chosen because of its simplicity

\section*{DATA ANALYSIS}




\section*{Findings}
- For all Olympics event, United State of America (USA) have won the competition 16 times as general champion. Further, Uni Soviet has 8 times as general champion
- Despite not being \(1^{\text {st }}\) position, USA also active as runner up and \(3^{\text {rd }}\) position
- Uni Soviet is a rival of USA
- German and Canada are the other rival of USA with good potency

\section*{DATA ANALYSIS}



\section*{Findings}
- As the rival of USA, Uni Soviet has the strongest sport with highest number of medals, that is wrestling
- The USA's sport with highest number of medals is athletics (28). It doesn't include in top ten sport won by the Uni Soviet
- Rowing, boxing, and diving can be optimized by USA in order to beat the real rival of Uni Soviet

\section*{DATA ANALYSIS}


Season has not effect on the performance of USA in Olympics event 1986-2016

\section*{DATA ANALYSIS}



Number of medal 1896-1932
\begin{tabular}{rrrrrrr} 
& Year & NOC & Bronze & Gold & Silver & All \\
\hline \(\mathbf{6}\) & 1896 & GRE & 20 & 10 & 18 & 48 \\
\(\mathbf{1 3}\) & 1900 & FRA & 82 & 52 & 101 & 235 \\
\(\mathbf{4 2}\) & 1904 & USA & 125 & 128 & 141 & 394 \\
\(\mathbf{5 2}\) & 1906 & GRE & 30 & 24 & 48 & 102 \\
\(\mathbf{6 7}\) & 1908 & GBR & 90 & 147 & 131 & 368 \\
\(\mathbf{9 7}\) & 1912 & SWE & 25 & 103 & 62 & 190 \\
\(\mathbf{1 0 8}\) & 1920 & USA & 38 & 111 & 45 & 194 \\
\(\mathbf{1 3 2}\) & 1924 & USA & 50 & 98 & 46 & 194 \\
\(\mathbf{1 6 5}\) & 1928 & USA & 19 & 53 & 30 & 102 \\
\(\mathbf{1 9 9}\) & 1932 & USA & 64 & 91 & 68 & 223
\end{tabular}

\section*{Findings}
- In 1980, it was a year with the highest number of medals to be contested
- Of course there is high positive correlation between number of sport with the total medals won by country (0.883)

\section*{CONCLUSION}
- The USA dominates the Olympics event as the top three with highest medal in 1986-2016
- To defeat the USA in Olympics, other country must be discipline in sports training, especially athletics, swimming, and wrestling
- Other country are recommended to gain the medal from the list of 15 sport that had never been won by the USA in 1986-2016
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