# Predicting the outcome of a UFC fight

Navin Sanjay April 2023

#### Purpose

 Predictive modelling was used to find the likelihood of a fighter winning by KO/Submission or by Judge Decision.

#### Target Audience:

- Betting agencies can adjust their odds accordingly
- Individuals can use this information when betting

#### Background

- The UFC (Ultimate Fighting Championship) is a mixed martial arts organization that features highly skilled fighters from around the world competing against each other in a variety of weight classes.
- Data was taken from the last 10 years of the Lightweight category and assessed to predict the likelihood of a match resulting in a KO/Submission or Judge Decision.
  - KO: Knockout
  - Submission: Fighter causes opponent to give up
  - Judge Decision: Outcome of the fight is determined by three judges scoring each round based on the fighters' performance.

# Data Overview and Data Pre-processing

#### Overview of Data

- Data was obtained from Kaggle
  - https://www.kaggle.com/datasets/fatismajli/ufc-data
- UFC Events Data
  - Data on each event and their results
  - 771 rows x 11 cols
- UFC Fighter Data
  - Data on each UFC fighter and their attributes
  - 1692 rows x 8 cols

### Overview of Data – UFC Fighter Data

	First Name	Last Name	Nickname	Height	Weight	Reach	Stance	Wins	Losses	Draws
0	Tom	Aaron	NaN		155 lbs.		NaN	5	3	0
1	Danny	Abbadi	The Assassin	5' 11"	155 lbs.		Orthodox	4	6	0
2	Nariman	Abbasov	Bayraktar	5' 8"	155 lbs.	66.0"	Orthodox	28	4	0
3	David	Abbott	Tank	6' 0"	265 lbs.		Switch	10	15	0
4	Hamdy	Abdelwa hab	The Hammer	6' 2"	264 lbs.	72.0"	Southpaw	5	0	0

#### Overview of Data — UFC Event Data

	<b>Event Name</b>	Event Date	Result	Fighter1	Fighter2	KD	Strikes	TD	Sub	Weight Class	Method	Round	Time
	UFC 286: Edwards vs. Usman 3	2023-03-18	Justin Gaethje	Justin Gaethje	Rafael Fiziev	0-0	103-97	1-0	0-0	Lightweight	M-DEC	3	5.000000
	UFC 286: Edwards vs. Usman 3	2023-03-18	Chris Duncan	Chris Duncan	Omar Morales	0-0	53-47	5-0	0-0	Lightweight	S-DEC	3	5.000000
17	UFC 286: Edwards vs. Usman 3	2023-03-18	Yanal Ashmouz	Yanal Ashmouz	Sam Patterson	0-0	15-2	1-0	0-0	Lightweight	KO/TKO- Punches	1	1.250000
23	UFC 286: Edwards vs. Usman 3	2023-03-18	Draw	Jai Herbert	Ludovit Klein	0-0	43-47	1-2	0-0	Lightweight	M-DEC	3	5.000000
	UFC 285: Jones vs. Gane	2023-03-04		Mateusz Gamrot	Jalin Turner	0-1	29-40	4-0	0-0	Lightweight	S-DEC	3	5.000000

#### Overview of Data – Data Cleaning

Object (string)

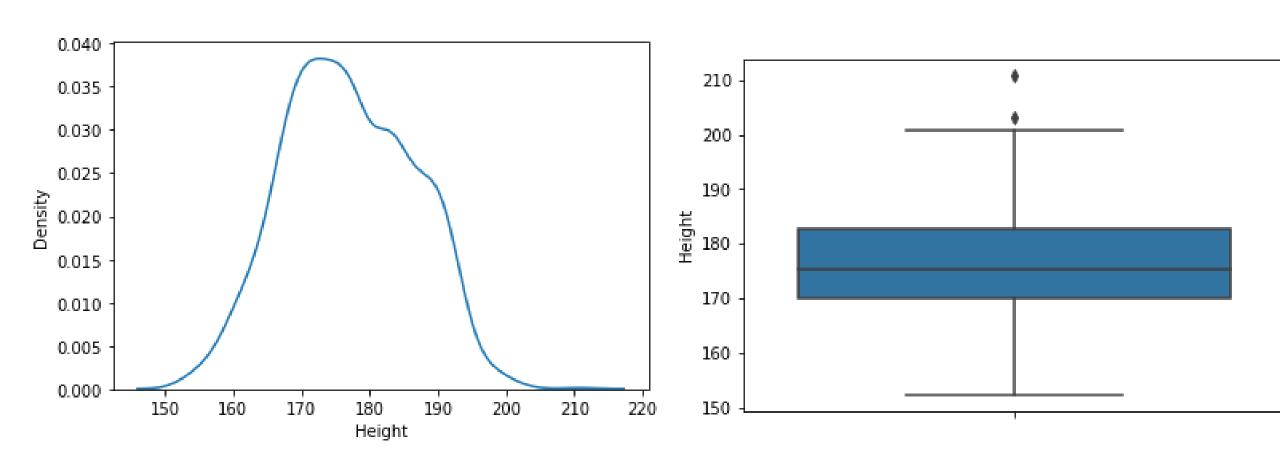
Integers & Floats

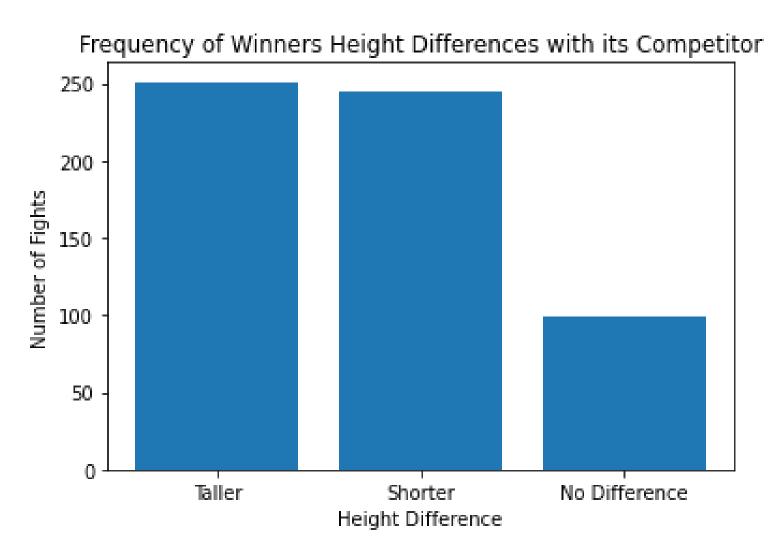
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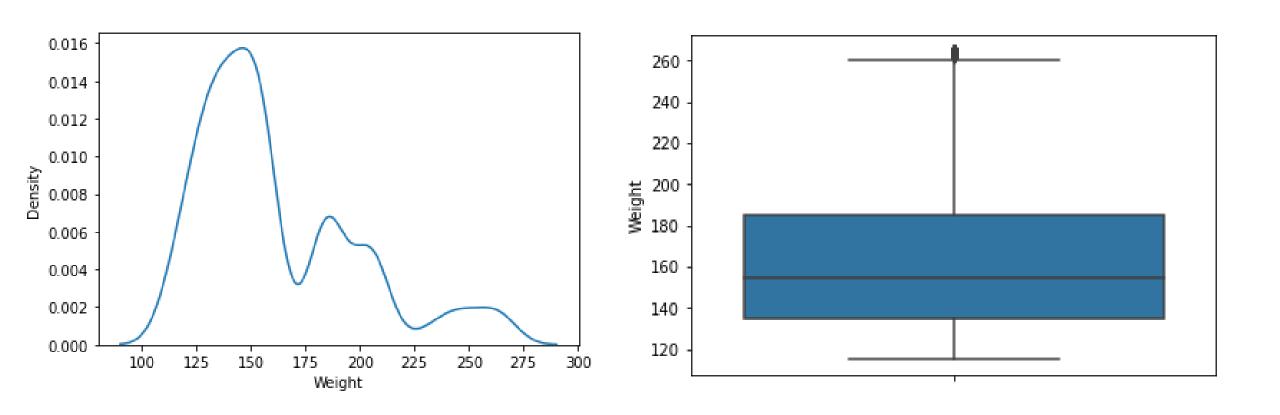
#### Overview of Data – Data Cleaning

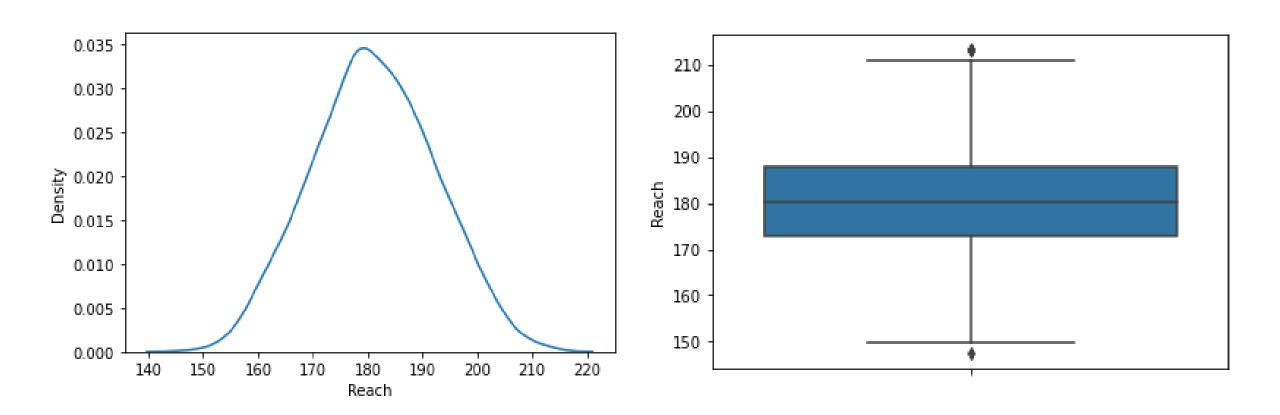
Target variable

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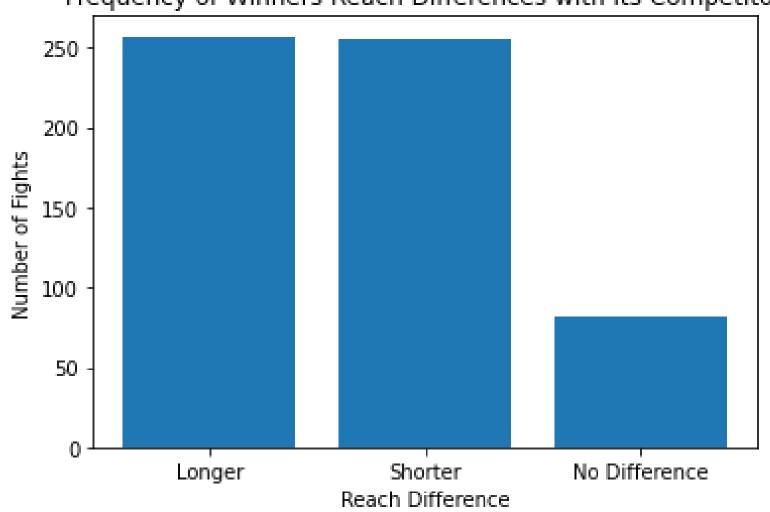


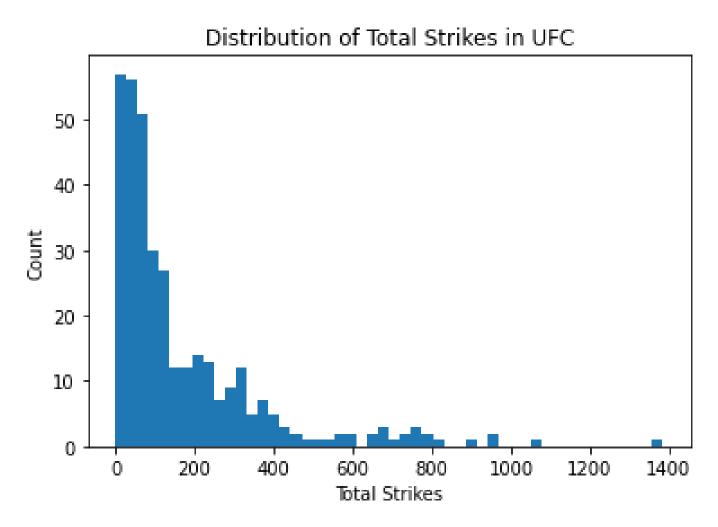


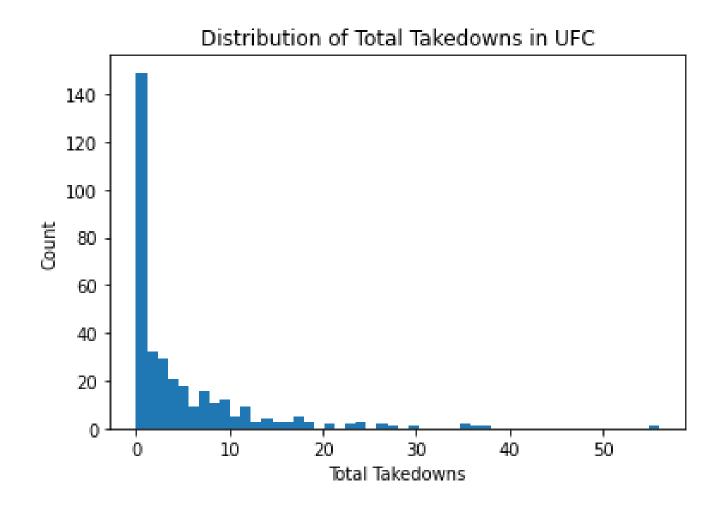


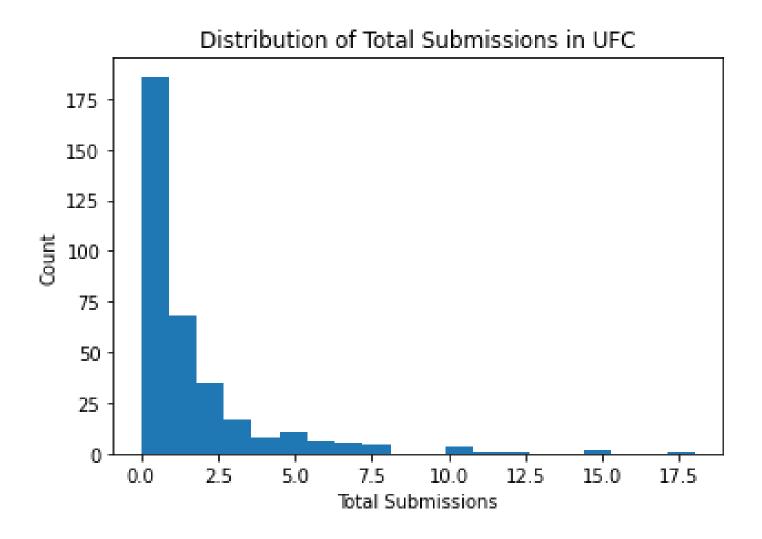


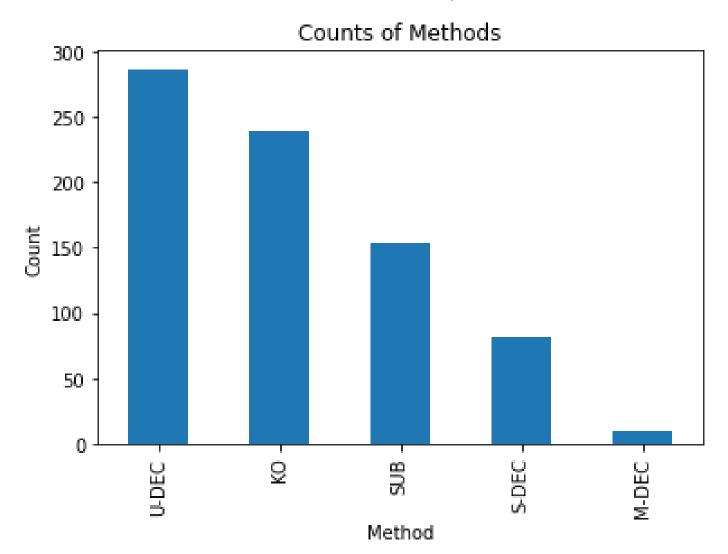
Frequency of Winners Reach Differences with its Competitor

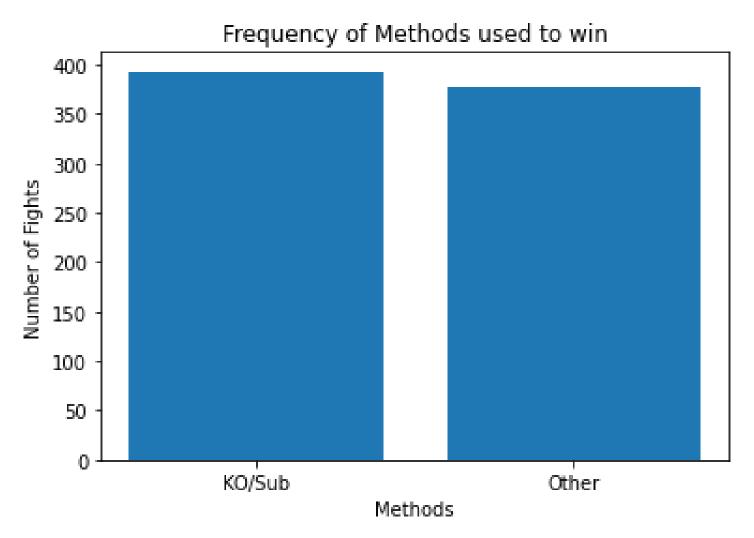












### Feature Selection

### Most relevant features for the Predictive Model

- Fighter1 and Fighter2 Match
  - Method of Win Predictor Variable
    - KO/Sub vs Anything Else
  - Total **Strike Difference** between fighters (For the last 10 years)
  - Total Takedown Difference
  - Total Submission Difference
  - Total Height Difference
  - Total Reach Difference
  - Total # Wins Difference
  - Total Experience Difference (Number of fights in the last 10 years)
  - Average Fight Time Difference
  - Total # KO wins Difference

Most relevant features for the Predictive

Model

	Method
Method	1.000000
strike_diff	-0.007704
td_diff	-0.027328
sub_diff	<mark>0.153585</mark>
Height_Diff	0.025767
Reach_Diff	0.045531
Wins_Diff	0.036651
Exp_Diff	0.029306
avg_fightti me_diff	<mark>-0.128979</mark>
ko_diff	0.058258

### Most relevant features for the Predictive Model

	strike_diff	td_diff	sub_diff	Height_Diff	Reach_Diff	Wins_Diff	Exp_Diff	avg_fighttime _diff	ko_diff
0	323	-1	0	7.62	-2.54	12	14	1.365720	5
1	-122	4	0	-2.54	-5.08	-1	-4	-2.208333	-1
2	13	1	0	-15.24	-25.40	-3	-6	-0.000000	0
3	41	24	-5	-12.70	-17.78	9	5	-5.352381	-1
4	-26	11	-1	2.54	0.00	6	10	-0.000000	0
589	146	-1	0	-2.54	-7.62	-26	-41	-5.145833	-2
590	-373	-9	3	-7.62	-2.54	-12	-19	0.505303	-5
591	-40	-1	0	0.00	10.16	5	0	4.300000	-1
592	561	55	7	7.62	0.00	8	0	-9.335000	3
593	729	4	1	-2.54	7.62	3	11	-8.144608	6

### Model Selection and Training

#### Different Models Used

- For the predictive modelling problem of classifying UFC fights as ending in a KO/Sub or by Decision, three popular binary classification algorithms were selected:
  - Logistic Regression
  - Support Vector Machine
  - Naïve Bayes (Gaussian)
- The performance of models was compared based on:
  - Accuracy Score
  - AUC ROC
  - F1-Score
- The models were trained on historical fight data (the last 10 years)

### Results and Analysis

### Model Comparison

Model	Accuracy Score	AUC – ROC	F1-Score
Logistic Regression	67.2%	69%	67%
Support Vector Machine	<mark>69.7%</mark>	<mark>71%</mark>	<mark>67%</mark>
Naive Bayes	63.9%	67%	61%

#### **Model Limitations**

#### Limited Data:

- My data was only on the last 10 years
- If there is few data of a certain fighter the model may not be able to make accurate predictions for those cases
- Combat Sport Unpredictability:
  - Combat sports are known for their unpredictability as it just takes one hit to knock someone out.
- Binary Classified:
  - In this project I only looked at whether it was a KO/Sub or Decision. However, it would be better to look at it as a multi-class problem.
- Interpretability:
  - SVM's (The model I chose as the best) are black box, therefore its difficult to interpret how the model is making its predictions

### Conclusion

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• In summary, I have demonstrated the power of predictive modelling in the context of UFC fights, showing how a combination of data preprocessing, feature engineering, and model selection can yield accurate predictions of fight outcomes.

 This approach has important implications not just for UFC but for other sports, where accurate predictions can inform everything from betting odds to training strategies.

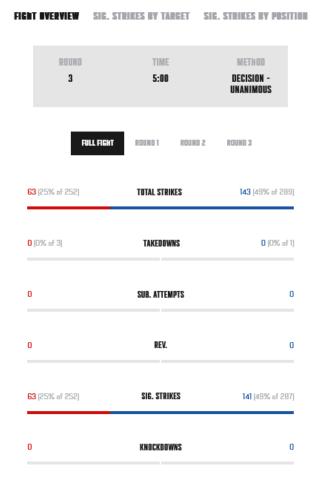
#### Conclusion

 While the model has shown promising results, there is always room for improvement, further exploration and refinement of predictive models can be done to further improve the accuracy.

 By continuing to refine our models and better understand the factors that influence fight outcomes, we can help fighters, coaches, fans and businesses alike make more informed decisions and gain deeper insights into the complex world of combat sports.

### Tested on recent Fight







#### Tested on recent Fight

- Fighter1 = Clay Gaudia
- Fighter2 = Rafa Garcia
- Model Predicted that the result would be "0"
  - Predicting that Fighter2 would win by Decision.

### Thank you!