

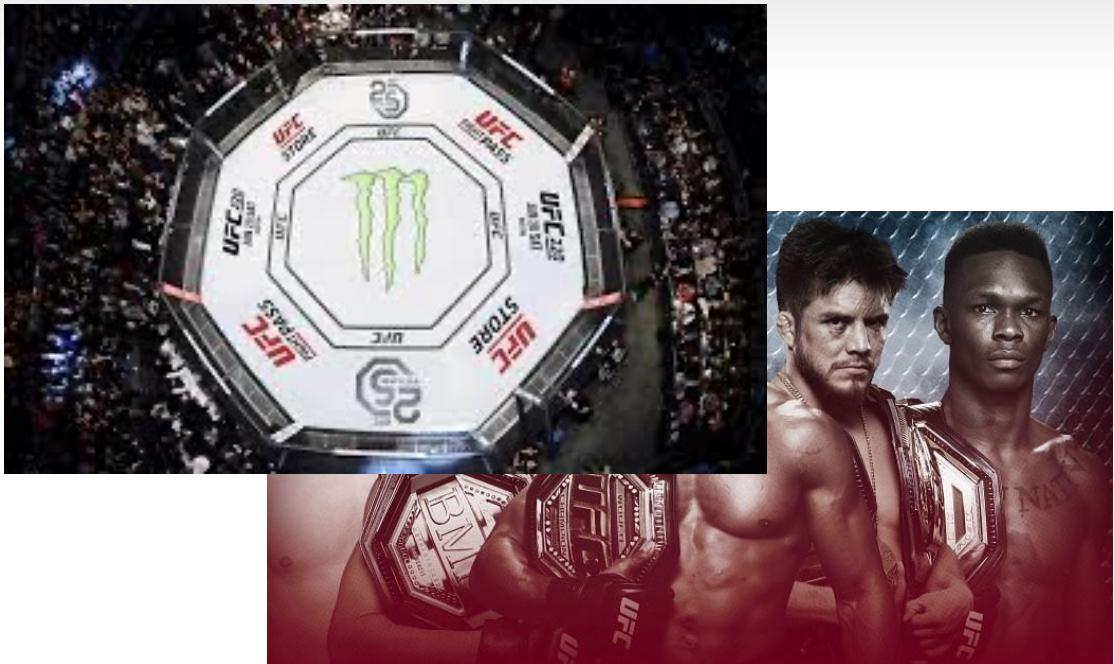
MMA STRIKE ANALYSIS USING COMPUTER VISION

GA DSI 25 – HAZIQ

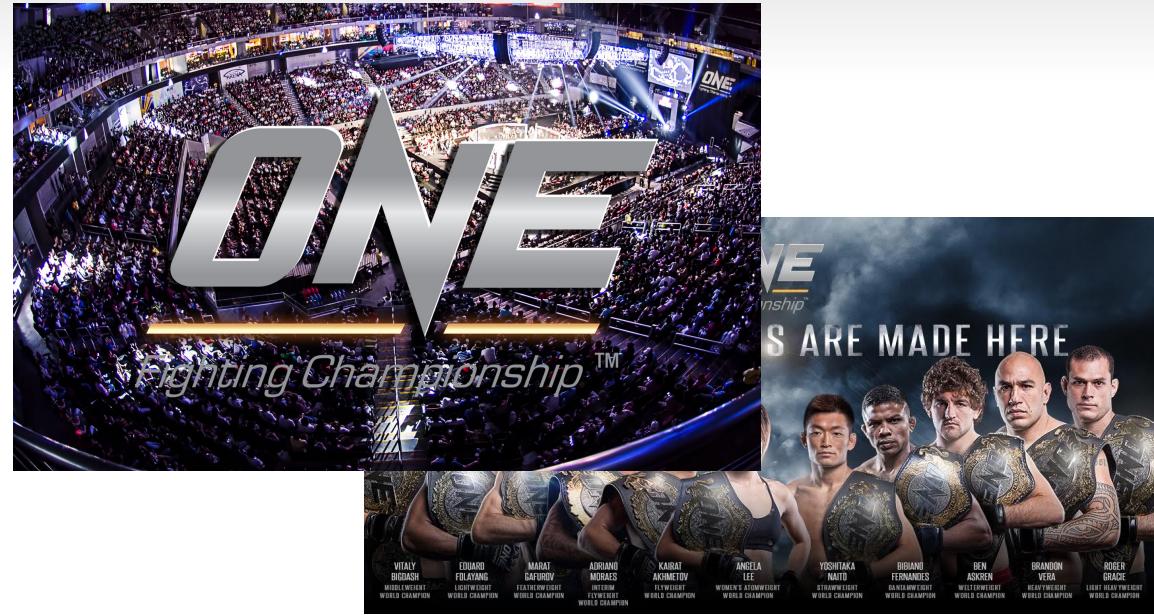
Overview

- Background
- Problem Statement
- Data Preprocessing
- EDA
- Modelling
- Future Improvements
- Conclusion

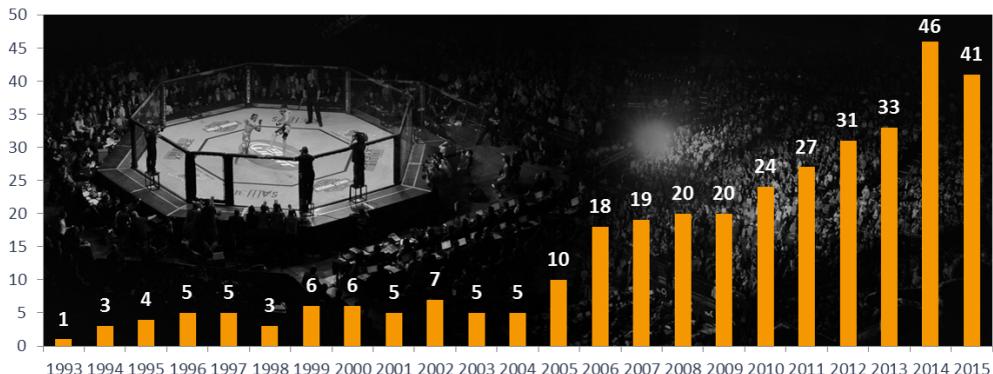
UFC: USD 7 billion



ONE Championship: USD 1 billion



Number of events



Background

A quick introduction on MMA

- Multi-disciplinary fighting styles:

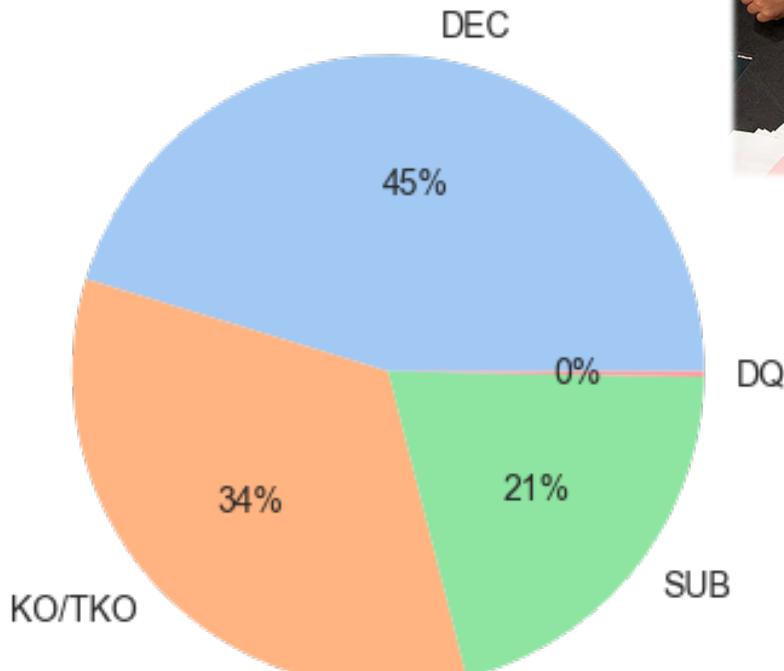
- Boxing
- Kickboxing
- Muay Thai
- Taekwondo
- Karate
- Brazilian Jiu-Jitsu (BJJ)
- Wrestling
- Sambo



Background

A quick introduction on MMA

- 3 methods for fights to end:
 - Decision
 - KO/TKO
 - Submission
- 45% of fights are decided by judges



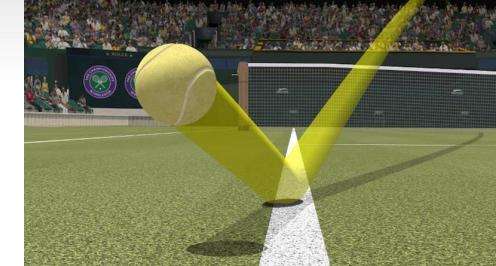
How are fights scored?

- Effective striking
 - Effective grappling
 - Effective aggressiveness
 - Control of fighting area
-
- Strikes attempted
 - Significant strikes landed
 - Striking accuracy

W/L	FIGHTER	KD	STR	TD	SUB	WEIGHT CLASS	METHOD	ROUND	TIME
WIN ➤	<u>Charles Oliveira</u>	0	73	0	3	Lightweight	SUB	3	1:02
	<u>Dustin Poirier</u>	1	58	0	0	 *PERF	Rear Naked Choke		
WIN ➤	<u>Julianne Pena</u>	0	79	1	1	Women's Bantamweight	SUB	2	3:26
	<u>Amanda Nunes</u>	0	46	1	0	 *PERF	Rear Naked Choke		
WIN ➤	<u>Geoff Neal</u>	0	85	0	0	Welterweight	 S-DEC	3	5:00
	<u>Santiago Ponzinibbio</u>	0	91	1	0				
WIN ➤	<u>Kai Kara-France</u>	2	23	0	0	Flyweight	KO/TKO	1	3:21
	<u>Cody Garbrandt</u>	0	9	0	0	 *PERF	Punches		
WIN ➤	<u>Sean O'Malley</u>	1	39	0	0	Bantamweight	KO/TKO	1	4:42
	<u>Raulian Paiva</u>	0	11	0	0	 *PERF	Punches		

Problem Statement

- **Problem:** Judging the winner of a fight can be tricky and even subjective at times.
- **Objective:** Assist judges in making their decision based on striking numbers.
- **Solution:** Computer vision tool to assist judges in scoring rounds and reduce subjectivity



Tennis – Hawk Eye



Football – Goal Line Technology



Rugby – Video Refereeing

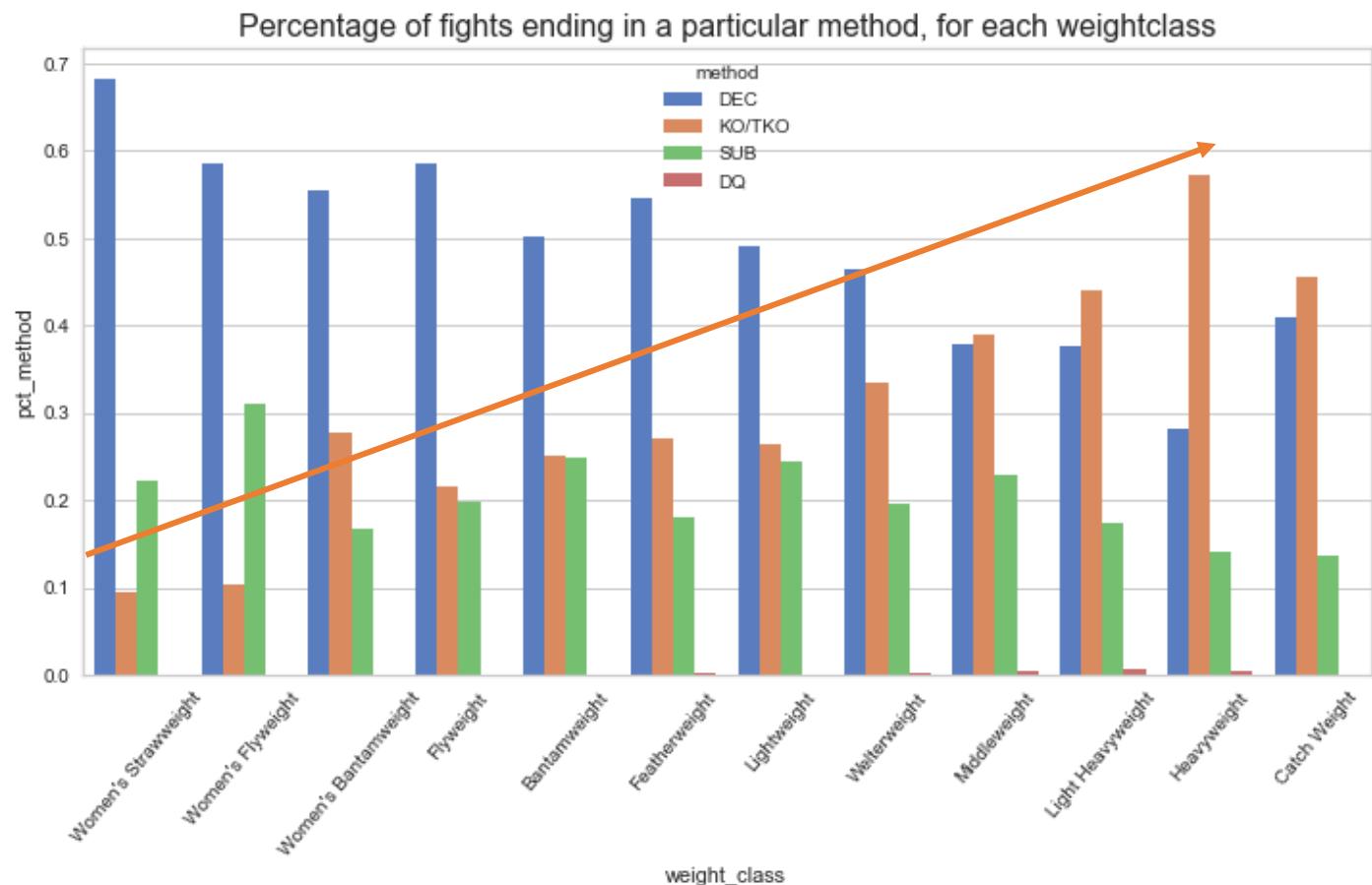
Data Cleaning (Feature Selection)

- Fighter-specific features
 - Age, Height, Reach, Stance, Weight Class
- Fight-related features
 - **Strikes:** Landed, Absorbed, Defended, Accuracy
 - Takedown: Landed, Defended, Accuracy
 - Submission: Attempted
- Other features
 - Round end, Method, Title Fight

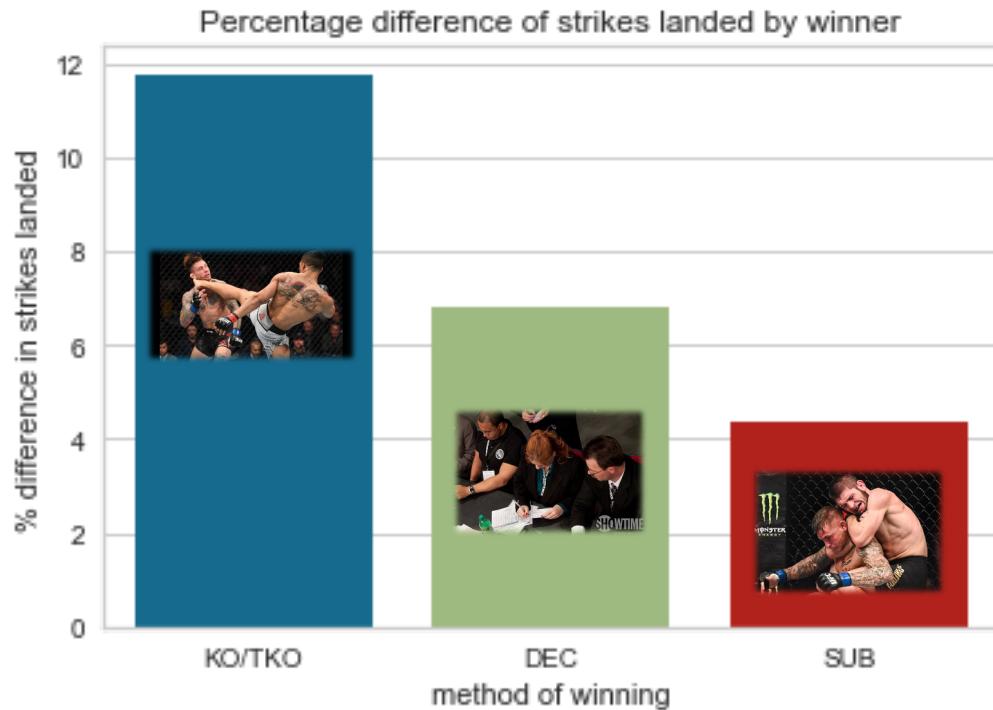
	fighter1	fighter2	winner	weight_class	title_fight	method	end_round	fight_year	win_fighter1	lose_fighter1	draw_fighter1	nc_fighter1	height_fighter1
4586	Chas Skelly	Bobby Moffett	0	Featherweight	f	SUB	2	2018	17	4	0	0	5' 11"
4587	Davi Ramos	John Gunther	1	Lightweight	f	SUB	1	2018	9	2	0	0	5' 6"
4588	Julian Erosa	Devonte Smith	0	Lightweight	f	KO/TKO	1	2018	22	7	0	0	6' 1"
4589	Eric Shelton	Joseph Morales	1	Flyweight	f	DEC	3	2018	12	5	0	0	5' 6"
4590	Mark De La Rosa	Joby Sanchez	1	Bantamweight	f	DEC	3	2018	11	1	0	0	5' 6"

EDA

- As weight class increases, higher percentage of fights end in KO/TKO



EDA



- Fighters that win by KO/TKO land more strikes against their opponents than those that win by SUBS

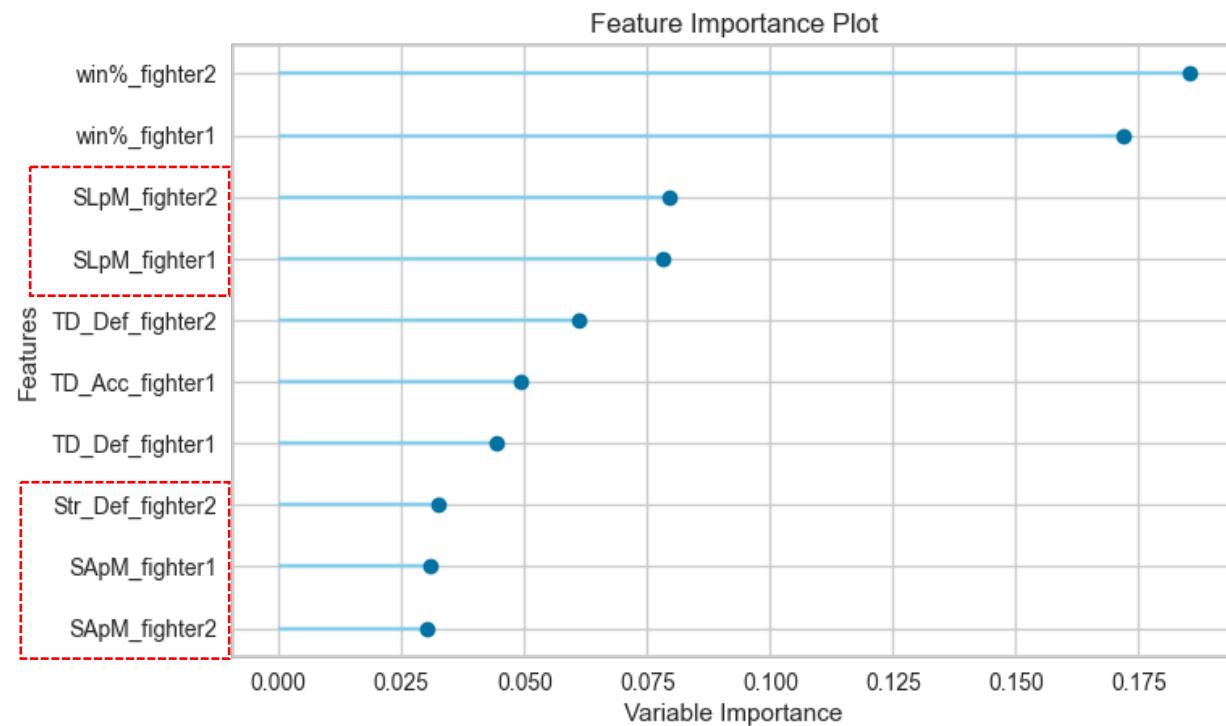
MMA Fight Predictor



Modelling

Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
gbc Gradient Boosting Classifier	0.7104	0.7857	0.7075	0.7118	0.7095	0.4207	0.4209	0.2620

- Accuracy on Test Set: 78%



- SLpM: Strikes Landed per Minute
- Str_Def : Strikes Defence
- SApM: Strikes Absorbed per Minute

Prediction

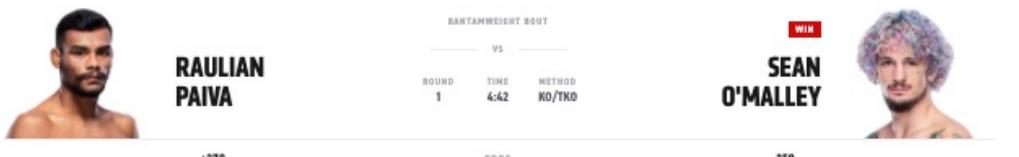
Predicted winner (83%)



Predicted winner (80%)



Predicted winner (80%)



Predicted winner (60%)

Predicted winner (65%)

BR BLEACHER REPORT Julianna Peña Beats Amanda Nunes via Submission to Win Title in UFC 269 Upset 2 days ago

ESPN The world doubted Julianna Peña could beat Amanda Nunes at UFC 269, but Peña believed 2 days ago

The Guardian UFC 269: Julianna Peña stuns Amanda Nunes in one of MMA's biggest upsets 1 day ago

More news →

Computer Vision Analysis of Strikes



Data Collection

- Video footage of MMA fights from online sources
- Crop out only the relevant fighter and apply pose estimation
- Extract pose estimation keypoints into dataframe



attack_type	frame_num	nose_x	nose_y	nose_z	left_eye_inner_x	left_eye_inner_y	left_eye_inner_z	left_eye_x	left_eye_y	left_eye_z	left_eye_outer
0	0	0.578093	0.122442	-0.040693	0.590232	0.118591	-0.033230	0.593823	0.122611	-0.033219	0.59754
1	0	1.0582941	0.120818	-0.051595	0.594627	0.118439	-0.044298	0.597930	0.122459	-0.044334	0.60135
2	0	2.0584567	0.121887	-0.066308	0.595633	0.119053	-0.059041	0.599057	0.122886	-0.059042	0.60261
3	0	3.0584130	0.121001	-0.076620	0.595297	0.118088	-0.066500	0.598840	0.121771	-0.066535	0.60254
4	0	4.0584493	0.121604	-0.066044	0.596268	0.118704	-0.055056	0.599892	0.122532	-0.055091	0.60366
...
3451	5	19.0538360	0.096930	0.119413	0.535498	0.083333	0.081952	0.532861	0.082931	0.081933	0.53043
3452	5	20.0537787	0.097246	0.117007	0.535080	0.083731	0.079740	0.532514	0.083370	0.079726	0.53015
3453	5	21.0537535	0.097313	0.120358	0.534762	0.083770	0.083794	0.532224	0.083438	0.083782	0.52985
3454	5	22.0537890	0.097986	0.129107	0.535075	0.084356	0.092177	0.532517	0.084020	0.092178	0.53015
3455	5	23.0538068	0.098495	0.133983	0.535200	0.084829	0.096888	0.532638	0.084514	0.096893	0.53027

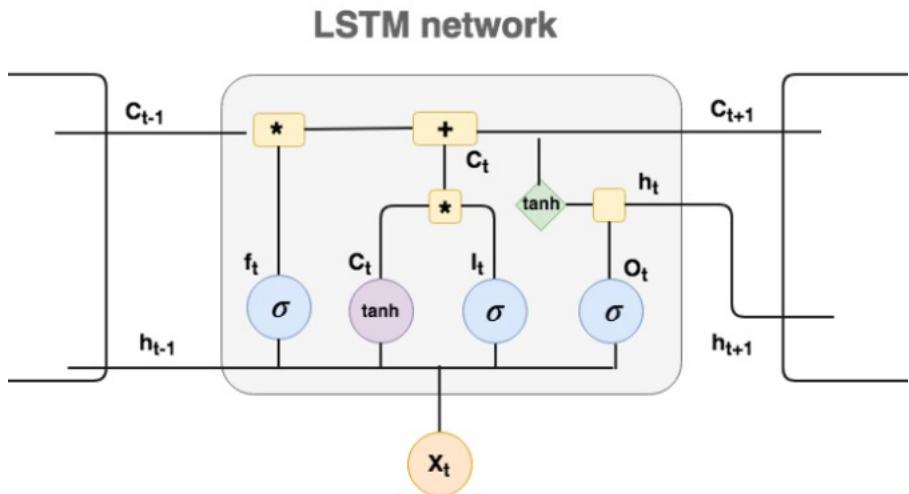


MediaPipe



Modelling (LSTM)

		attack_type	frame_num	nose_x	nose_y	nose_z	left_eye_inner_x	left_eye_inner_y	left_eye_inner_z	left_eye_x	left_eye_y	left_eye_z	left_eye_outer_x
0	0	0	0.578093	0.122442	-0.040693	0.590232	0.118591	-0.033230	0.593823	0.122611	-0.033219	0.59754	
1	0	1	0.582941	0.120818	-0.051595	0.594627	0.118439	-0.044298	0.597939	0.122459	-0.044334	0.6013E	
2	0	2	0.584567	0.121887	-0.068308	0.595633	0.119053	-0.059041	0.599057	0.122886	-0.059042	0.6026E	
3	0	3	0.584130	0.121001	-0.076620	0.595297	0.118088	-0.068500	0.598848	0.121771	-0.066535	0.6025E	
4	0	4	0.584493	0.121604	-0.066044	0.596268	0.118704	-0.055056	0.599892	0.122532	-0.055091	0.6036E	
...	
3451	5	19	0.538360	0.096936	0.119413	0.535498	0.083333	0.081950	0.532861	0.082931	0.081933	0.5304E	
3452	5	20	0.537787	0.097246	0.117007	0.535080	0.083731	0.079740	0.532514	0.083370	0.079726	0.5301E	
3453	5	21	0.537595	0.097313	0.120358	0.534762	0.083770	0.083794	0.532224	0.084348	0.083782	0.5298E	
3454	5	22	0.537890	0.097986	0.129107	0.535075	0.084356	0.092177	0.532517	0.084020	0.092178	0.5301E	
3455	5	23	0.538068	0.098495	0.133983	0.535200	0.084829	0.096888	0.532638	0.084514	0.096893	0.5302E	



- Train set

```

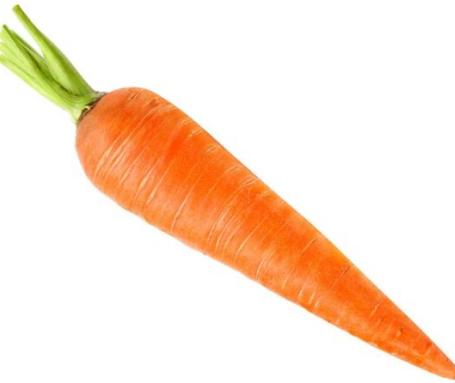
Epoch 200/200
34/34 - 1s - loss: 0.4114 - binary_accuracy: 0.8088 - accuracy: 0.7941
- val_loss: 0.7596 - val_binary_accuracy: 0.5000 - val_accuracy: 0.4667
  
```

- Test set

```

2/2 [=====] - 0s 10ms/step
- loss: 0.8901 - binary_accuracy: 0.5238 - accuracy: 0.5238
  
```

attack_type	frame_num	nose_x	nose_y	nose_z	left_eye_inner_x	left_eye_inner_y	left_eye_inner_z	left_eye_x	left_eye_y	left_eye_z	left_eye_outer_
0	0	0	0.578093	0.122442	-0.040693	0.590232	0.118591	-0.033230	0.593823	0.122611	-0.033219
1	0	1	0.582941	0.120818	-0.051595	0.594627	0.118439	-0.044298	0.597939	0.122459	-0.044334
2	0	2	0.584567	0.121887	-0.066308	0.595633	0.119053	-0.059041	0.599057	0.122886	-0.059042
3	0	3	0.584130	0.121001	-0.076620	0.595297	0.118088	-0.066506	0.598848	0.121771	-0.066535
4	0	4	0.584493	0.121604	-0.066044	0.596268	0.118704	-0.055056	0.599892	0.122532	-0.055091
...
3451	5	19	0.538360	0.096936	0.119413	0.535498	0.083333	0.081952	0.532861	0.082931	0.081933
3452	5	20	0.537787	0.097246	0.117007	0.535080	0.083731	0.079740	0.532514	0.083370	0.079726
3453	5	21	0.537535	0.097313	0.120358	0.534762	0.083770	0.083794	0.532224	0.083438	0.083782
3454	5	22	0.537890	0.097986	0.129107	0.535075	0.084356	0.092177	0.532517	0.084020	0.092178
3455	5	23	0.538068	0.098495	0.133983	0.535200	0.084829	0.096888	0.532638	0.084514	0.096893



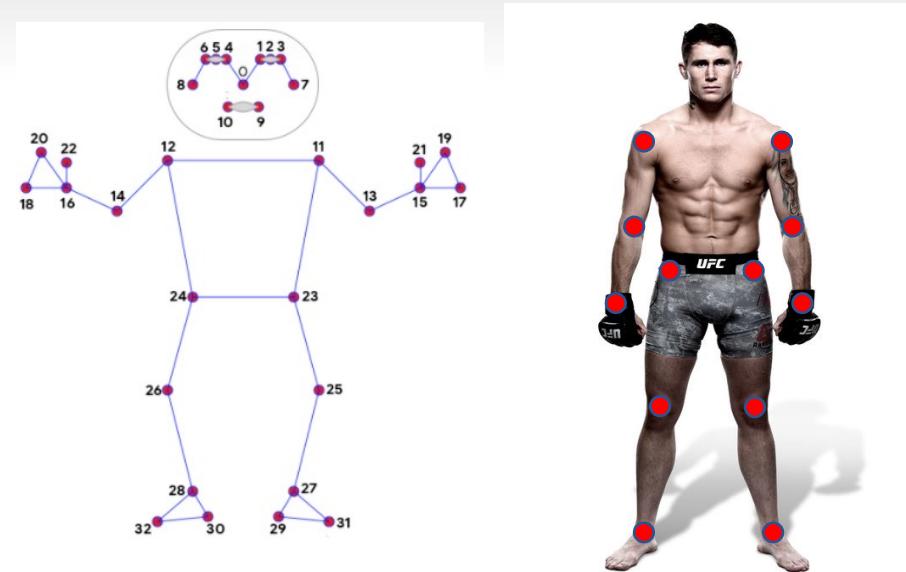
Extra Trees Classifier
Accuracy = 90%

Data Preprocessing

- Remove frames before a strike.
- 10 joints angles kept, other keypoints removed.
- Only still images of each strike type were used.
 - Total of 6 different strikes

attack_type	right_elbow_angle	right_shoulder_angle	right_hip_angle	right_knee_angle	right_ankle_angle	left_elbow_angle	left_shoulder_angle	
0	0	170.0	17.0	65.0	170.0	143.0	7.0	35.0
1	0	168.0	15.0	66.0	173.0	141.0	9.0	36.0
2	0	155.0	11.0	68.0	172.0	142.0	8.0	33.0
3	0	180.0	14.0	68.0	175.0	147.0	8.0	31.0
4	0	156.0	16.0	67.0	175.0	146.0	10.0	33.0
...	
3451	5	39.0	49.0	153.0	174.0	118.0	147.0	72.0
3452	5	36.0	49.0	153.0	174.0	119.0	148.0	73.0
3453	5	37.0	49.0	154.0	174.0	118.0	150.0	74.0
3454	5	35.0	49.0	154.0	174.0	117.0	150.0	74.0
3455	5	35.0	49.0	154.0	174.0	117.0	152.0	75.0

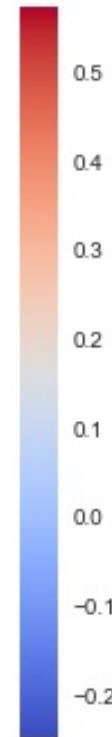
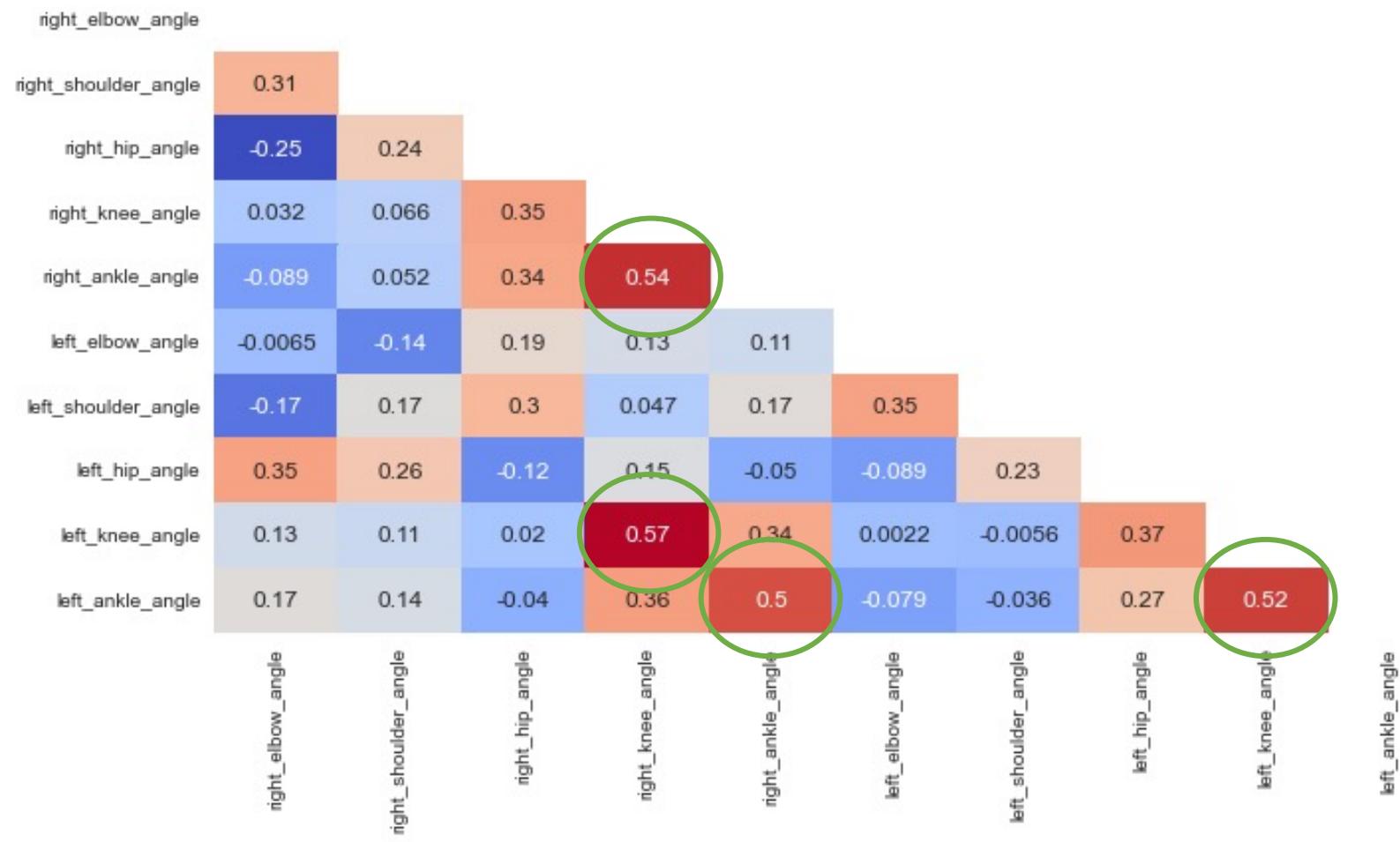
3456 rows × 11 columns



Collecting frames for right_highkick Video Number 6



EDA



Modelling

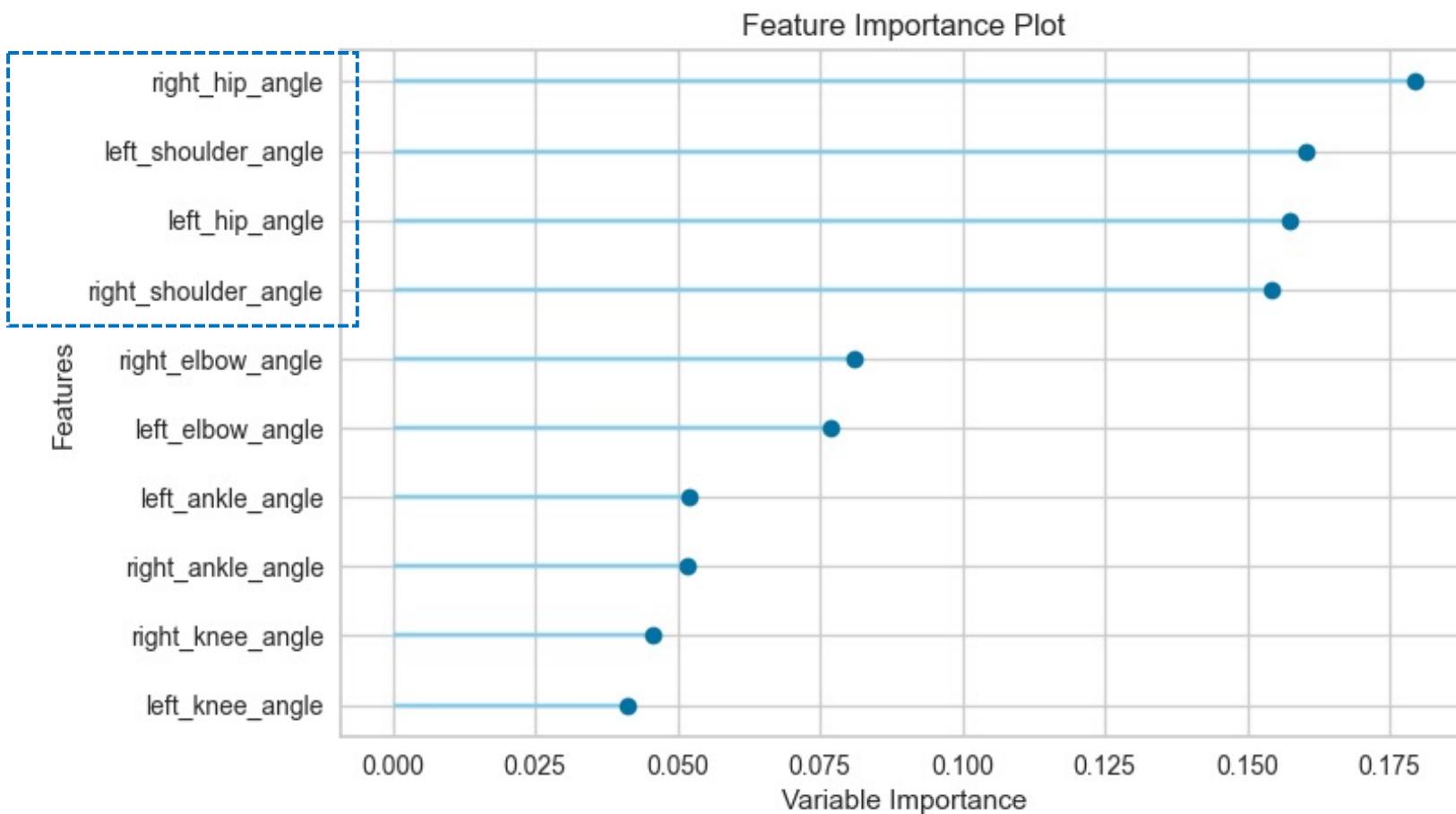
Train set

Model		Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
et	Extra Trees Classifier	0.9512	0.9968	0.9497	0.9514	0.9508	0.9414	0.9416	0.1710
xgboost	Extreme Gradient Boosting	0.9446	0.9960	0.9431	0.9460	0.9443	0.9335	0.9339	1.3920
lightgbm	Light Gradient Boosting Machine	0.9442	0.9965	0.9429	0.9461	0.9440	0.9330	0.9335	3.9870
rf	Random Forest Classifier	0.9438	0.9959	0.9424	0.9441	0.9434	0.9325	0.9327	0.2000
knn	K Neighbors Classifier	0.9351	0.9873	0.9337	0.9359	0.9349	0.9221	0.9223	0.0500
gbc	Gradient Boosting Classifier	0.9330	0.9935	0.9314	0.9336	0.9327	0.9196	0.9199	0.4660
dt	Decision Tree Classifier	0.9190	0.9538	0.9172	0.9199	0.9185	0.9027	0.9031	0.0060
qda	Quadratic Discriminant Analysis	0.8809	0.9739	0.8791	0.8846	0.8809	0.8571	0.8579	0.0050
lr	Logistic Regression	0.8499	0.9676	0.8471	0.8522	0.8495	0.8199	0.8205	0.2740
lda	Linear Discriminant Analysis	0.8272	0.9595	0.8260	0.8311	0.8246	0.7926	0.7943	0.0050
nb	Naive Bayes	0.8214	0.9496	0.8200	0.8246	0.8182	0.7857	0.7875	0.0050
ridge	Ridge Classifier	0.8127	0.0000	0.8120	0.8200	0.8111	0.7753	0.7773	0.0050
svm	SVM - Linear Kernel	0.7995	0.0000	0.7964	0.8133	0.7973	0.7594	0.7629	0.0240
ada	Ada Boost Classifier	0.6292	0.8582	0.6253	0.6332	0.6071	0.5546	0.5721	0.0300
dummy	Dummy Classifier	0.1728	0.5000	0.1667	0.0299	0.0509	0.0000	0.0000	0.0040

Test set

Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	
0	Extra Trees Classifier	0.9344	0.9956	0.9380	0.9355	0.9345	0.9212	0.9214

Feature Importance



Visualization of shoulder and hip angles



Left high kick



Left low kick

Potential use

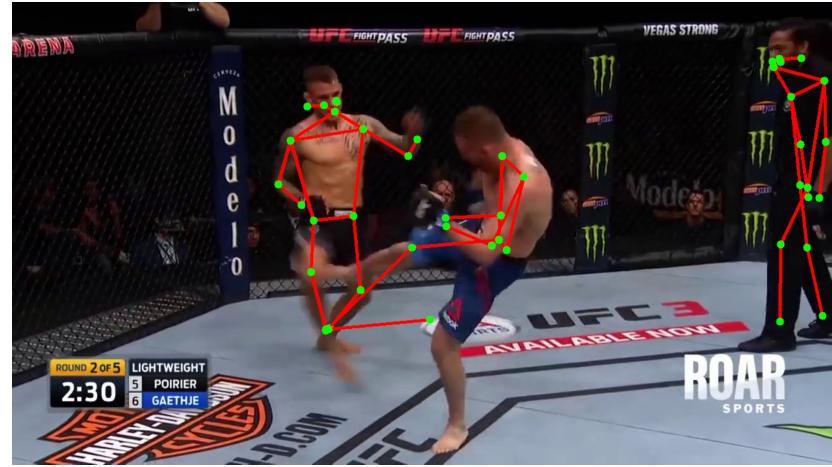


Potential use



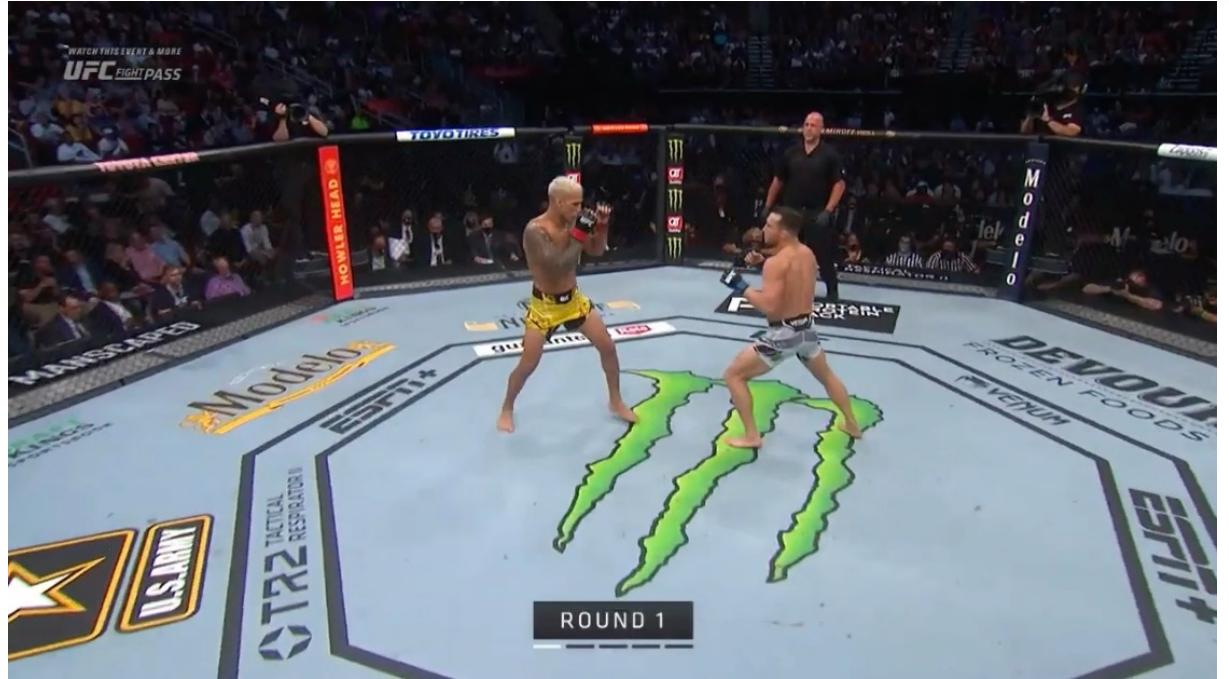
Challenges faced / Limitations

- No consistent camera angle.
- Unable to ‘fix’ the detection to one person.
- Editing each video is not feasible.



Future Improvements

- To include an ‘impact’ component
 - Minor, normal, significant
- Measuring other aspects
 - Ring control, aggressiveness
- Potential use to evaluate safety of fighter
 - Facial recognition.
 - Detect pain or lack of consciousness.



Conclusion

- Striking is a major aspect of MMA fights.
- 45% of fights are decided by judges.
- CV tool proposed to classify different types of strikes.
 - Accuracy = 93%
 - F1 = 93%
 - Assist MMA judges in scoring each round.



A dynamic photograph of two mixed martial arts fighters in a cage. One fighter, wearing black shorts, is in the foreground, leaning forward with his right arm raised in a fist. The other fighter, wearing grey shorts, is in the background, also leaning forward with his right arm raised. A referee in a black shirt is visible behind them. The background shows the dark interior of a cage with some branding.

THANK YOU