

F1 Rank Prediction Analysis

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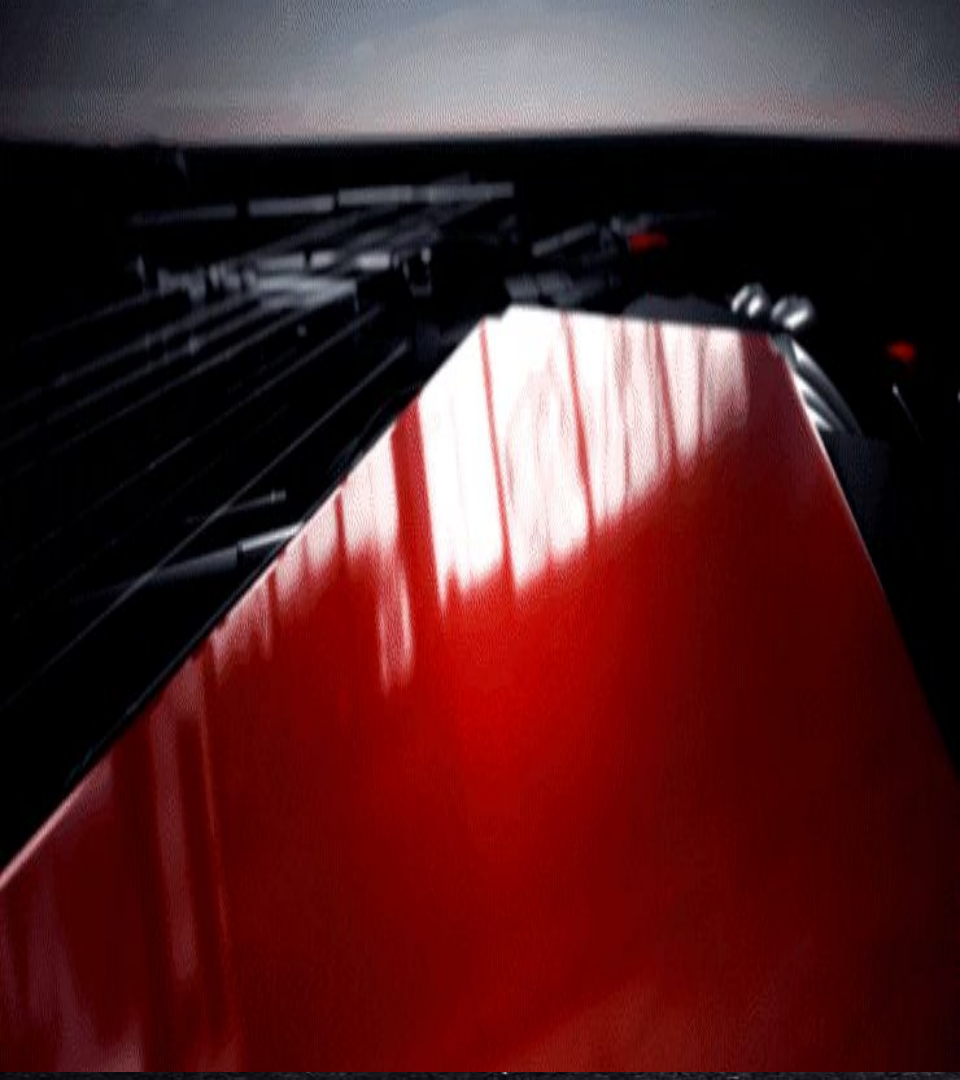


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Intro to F1



1. Circuits
2. Grid
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2019 FORMULA 1 SINGAPORE AIRLINES
SINGAPORE GRAND PRIX

BRAKES EFFORT **VERY HARD** TIME SPENT BRAKING **25%**

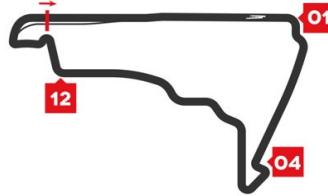


THREE HARDEST BRAKES

	Maximum deceleration	Max pedal load	Time spent braking
01	5.2 g	114 Kg	1.70 s
07	5.4 g	144 Kg	2.06 s
14	4.9 g	139 Kg	2.22 s

2019 FORMULA 1
GRAN PREMIO DE MÉXICO

BRAKES EFFORT **VERY HARD** TIME SPENT BRAKING **20%**



THREE HARDEST BRAKES

	Maximum deceleration	Max pedal load	Time spent braking
01	5.4 g	177 Kg	2.52 s
04	5.3 g	166 Kg	2.43 s
12	4.6 g	114 Kg	1.93 s

DESIGNED BY THE VISUAL AGENCY

2019 FORMULA 1
AZERBAIJAN GRAND PRIX

BRAKES EFFORT **HARD** TIME SPENT BRAKING **19%**



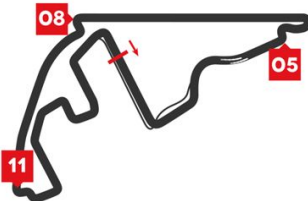
THREE HARDEST BRAKES

	Maximum deceleration	Max pedal load	Time spent braking
01	5.5 g	131 Kg	1.84 s
03	4.7 g	141 Kg	2.31 s
15	4.7 g	161 Kg	2.67 s

DESIGNED BY THE VISUAL AGENCY

2019 FORMULA 1 ETIHAD AIRWAYS
ABU DHABI GRAND PRIX

BRAKES EFFORT **VERY HARD** TIME SPENT BRAKING **19%**



THREE HARDEST BRAKES

	Maximum deceleration	Max pedal load	Time spent braking
05	5.2 g	117 Kg	1.76 s
08	5.6 g	186 Kg	2.57 s
11	5.0 g	155 Kg	2.38 s

DESIGNED BY THE VISUAL AGENCY

Circuits

68 Circuits

22 Grand Prix every season Ex:
Singapore, Mexico, Abu Dhabi,
Azerbaijan etc.

Source: <https://www.brembo.com/en/company/news/brembo-brake-analysis-abu-dhabi-formula-1>



Home Advantage



Max Verstappen fans in the Dutch Grand Prix



Ferrari fans in the Italian Grand Prix





Grid

20 Drivers

20 positions

Pole position

Qualifying race



Points system



F1 2022 DRIVER STANDINGS AFTER JAPANESE GRAND PRIX			F1 2022 TEAM STANDINGS AFTER JAPANESE GRAND PRIX		
	VERSTAPPEN	366		RED BULL RACING	619
	PEREZ	253		FERRARI	454
	LECLERC	252		MERCEDES-AMG	387
	RUSSELL	207		ALPINE	143
	SAINZ	202		MCLAREN	130
	HAMILTON	180		ALFA ROMEO	52
	NORRIS	101		ASTON MARTIN	45
	OCOON	78		HAAS	34
	ALONSO	65		ALPHATAURI	34
	BOTTAS	46		WILLIAMS	8
Positions			Points		
1st Position			25 Points		
2nd Position			18 Points		
3rd Position			15 Points		
4th Position			12 Points		
5th Position			10 Points		
6th Position			8 Points		
7th Position			6 Points		
8th Position			4 Points		
9th Position			2 Points		
10th Position			1 Point		

Podium

Top 3 drivers in a race get the podium

Most Podiums - Lewis Hamilton



F1 Season, 2022

Driver Championship: Max Verstappen



Constructors Championship: Redbull



Data

- Data source: Formula 1 World Championship from 1950 till the latest 2021 season (Source: Kaggle)
- Races, drivers, constructors, qualifying, circuits, lap times, pit stops, championships
- Data cleaning and feature engineering: Race results, lap times, previous performances, home advantage etc. to predict **positions**.
- Most relevant feature in prediction: Win rate, past avg pos, Constructor home advantage



Our Model - Feature Selection

OLS Regression Results

Dep. Variable:	podium	R-squared:	0.444
Model:	OLS	Adj. R-squared:	0.414
Method:	Least Squares	F-statistic:	3881.
Date:	Tue, 04 Dec 2022	Prob (F-statistic):	0.00
Time:	00:44:23	Log-likelihood:	-19336.
No. Observations:	14599	AIC:	7.487e+04
DF Residuals:	14595	BIC:	7.470e+04
DF Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Intercept	1.3499	0.006	14.398	0.000	1.344	1.354
win_rate	-1.0561	0.239	-4.435	0.000	-1.574	-0.438
Past_n_year_avg_pos	0.4363	0.010	47.149	0.000	0.417	0.456
constructor_home_adv	0.2083	0.108	1.944	0.049	0.001	0.416

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

Modelling and Scoring

- Implement Multiple ML Algorithms
- Getting in-sample and out-of-sample RMSE
- Improvement in R2 or adjusted R2 value

F1 Race Prediction

Data Manipulation and Sport Understanding

- Creating meaningful features by data manipulation
- Adding features contingent on performance and conditions

Correlation Analysis & Statistical Significance

- Checking for multicollinearity
- Checking for statistical significance

	date	year	round	constructorId	driverId	Past_n_year_race	points	position	win_rate	Past_n_year_avg_pos	constructor_home_adv
0	2008-03-16	2008	18	5	1	16	10.0	1	0.250000	3.000000	0
1	2008-03-16	2008	18	8	2	15	8.0	2	0.000000	5.333333	0
2	2008-03-16	2008	18	4	3	14	6.0	3	0.000000	9.142857	0
3	2008-03-16	2008	18	4	4	16	5.0	4	0.250000	2.937500	0
4	2008-03-16	2008	18	5	5	16	4.0	5	0.000000	7.875000	0
...
14594	2022-03-20	2022	1074	5	846	21	0.0	15	0.000000	6.619048	0
14595	2022-03-20	2022	1074	9	849	19	0.0	16	0.000000	15.105263	0
14596	2022-03-20	2022	1074	7	807	0	0.0	17	0.000000	10.000000	0
14597	2022-03-20	2022	1074	3	815	20	0.0	18	0.050000	6.600000	0
14598	2022-03-20	2022	1074	3	830	19	0.0	19	0.526316	1.842105	0

14599 rows x 11 columns

	Past_n_year_podium_driver	Past_n_year_wins	Past_n_year_pole	points	win_rate	podium_rate	points_rate	pole_rate	winner
Past_n_year_podium_driver	1.000000	0.845604	0.817489	0.749963	0.847912	0.985815	0.944407	0.815696	0.496339
Past_n_year_wins	0.845604	1.000000	0.936935	0.616576	0.992022	0.811022	0.784615	0.923964	0.533398
Past_n_year_pole	0.817489	0.936935	1.000000	0.597625	0.926986	0.782466	0.750097	0.992328	0.522727
points	0.749963	0.616576	0.597625	1.000000	0.830430	0.754921	0.788201	0.604642	0.641595
win_rate	0.847912	0.992022	0.926986	0.830430	1.000000	0.830729	0.802294	0.927662	0.540928
podium_rate	0.985815	0.811022	0.782466	0.754921	0.830729	1.000000	0.953083	0.796467	0.490391
points_rate	0.944407	0.784615	0.750097	0.788201	0.802294	0.953083	1.000000	0.760986	0.471896
pole_rate	0.815696	0.923964	0.992328	0.604642	0.927662	0.796467	0.760986	1.000000	0.524595
winner	0.496339	0.533398	0.522727	0.641595	0.540928	0.490391	0.471896	0.524595	1.000000



Final Data Set

	date	year	raceId	constructorId	driverId	Past_n_year_races	points	position	win_rate	Past_n_year_avg_pos	constructor_home_adv	
0	2008-03-16	2008	18	5	1	16	10.0	1	0.250000	3.000000	0	
1	2008-03-16	2008	18	8	2	15	8.0	2	0.000000	5.333333	0	
2	2008-03-16	2008	18	4	3	14	6.0	3	0.000000	9.142857	0	
3	2008-03-16	2008	18	4	4	16	5.0	4	0.250000	2.937500	0	
4	2008-03-16	2008	18	5	5	16	4.0	5	0.000000	7.875000	0	
...	
14594	2022-03-20	2022	1074	5	846	21	0.0	15	0.000000	6.619048	0	
14595	2022-03-20	2022	1074	9	849	19	0.0	16	0.000000	15.105263	0	
14596	2022-03-20	2022	1074	7	807	0	0.0	17	0.000000	10.000000	0	
14597	2022-03-20	2022	1074	3	815	20	0.0	18	0.050000	6.600000	0	
14598	2022-03-20	2022	1074	3	830	19	0.0	19	0.526316	1.842105	0	

14599 rows x 11 columns

Model Selection

Random Model - Made a model to randomly rank drivers in each race

Train (2000 - first 5 races of 2021) and Test set (remaining races of 2021)

Model	Z Score Data Set	Absolute Value Data Set
Random Model	5.86	5.86
Linear Regression	2.70	2.70
Random Forest	3.54	3.54
KNN	3.57	3.57

P-value: **Linear regression** (2.78×10^{-28}) > Random Forest > KNN (1.6×10^{-26})



Final Model Summary



OLS Regression Results

```
=====
Dep. Variable:          position    R-squared:                0.444
Model:                  OLS        Adj. R-squared:            0.444
Method:                 Least Squares  F-statistic:             3881.
Date:                   Tue, 06 Dec 2022  Prob (F-statistic):      0.00
Time:                   00:46:23      Log-Likelihood:          -39330.
No. Observations:      14599        AIC:                    7.867e+04
Df Residuals:          14595        BIC:                    7.870e+04
Df Model:               3
Covariance Type:       nonrobust
=====
```

```
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept          1.3499      0.094      14.398      0.000        1.166        1.534
win_rate          -1.1061      0.239      -4.635      0.000       -1.574       -0.638
Past_n_year_avg_pos  0.8363      0.010     85.168      0.000        0.817        0.856
constructor_home_adv  0.2083      0.106       1.966      0.049        0.001        0.416
=====
```

```
=====
Omnibus:            1020.288    Durbin-Watson:           1.069
Prob(Omnibus):      0.000      Jarque-Bera (JB):        1523.579
Skew:               0.577      Prob(JB):                0.00
Kurtosis:           4.082      Cond. No.                73.4
=====
```

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

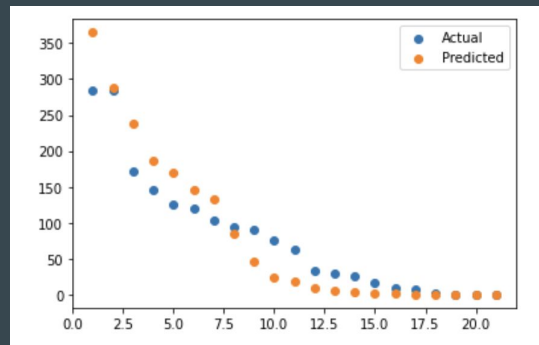


2021 Season Prediction Comparison

Pos	Our Prediction	Original Results
1	M. Verstappen	M. Verstappen
2	L. Hamilton	L. Hamilton
3	V. Bottas	V. Bottas
4	C. Sainz Jr	S. Perez
5	L. Norris	C. Sainz Jr
6	S. Perez	L. Norris
7	C. Leclerc	C. Leclerc
8	D. Ricciardo	D. Ricciardo
9	P. Gasly	P. Gasly
10	E. Ocon	F. Alonso

Prediction Accuracy: 80%

S. Perez changed teams from Racing Point to Redbull. This gave him a car advantage resulting in a better performance compared to his previous season, giving him Position 4 in original results and Position 6 in our prediction.





Thank You