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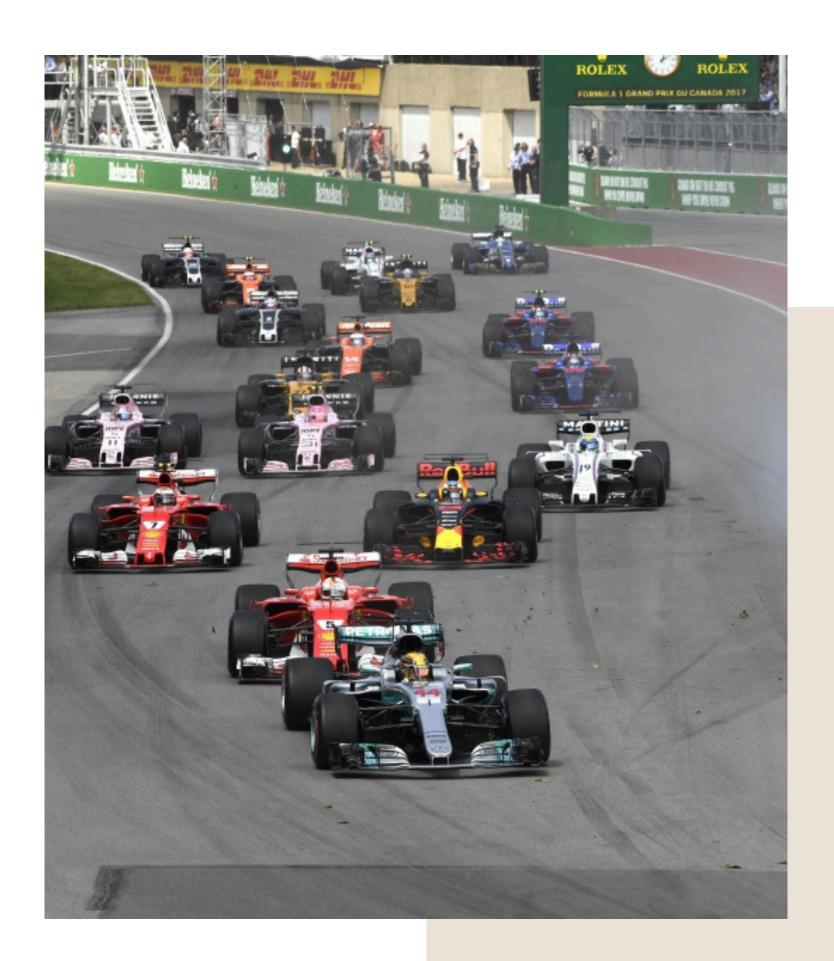
Analysis of the pit crew performance and its relationship to the Formula 1 championship standings

Actuarial Data Science



Introduction

- Formula 1 is a sport of precision, where the cars are **battling for every tenth of seconds**
- During a race, cars spend dozens of seconds in the pit lane
- Car performance and drivers' skills are often commented
- For this presentation, we will look at the importance of the pit stop efficiency on overall performance in Formula 1



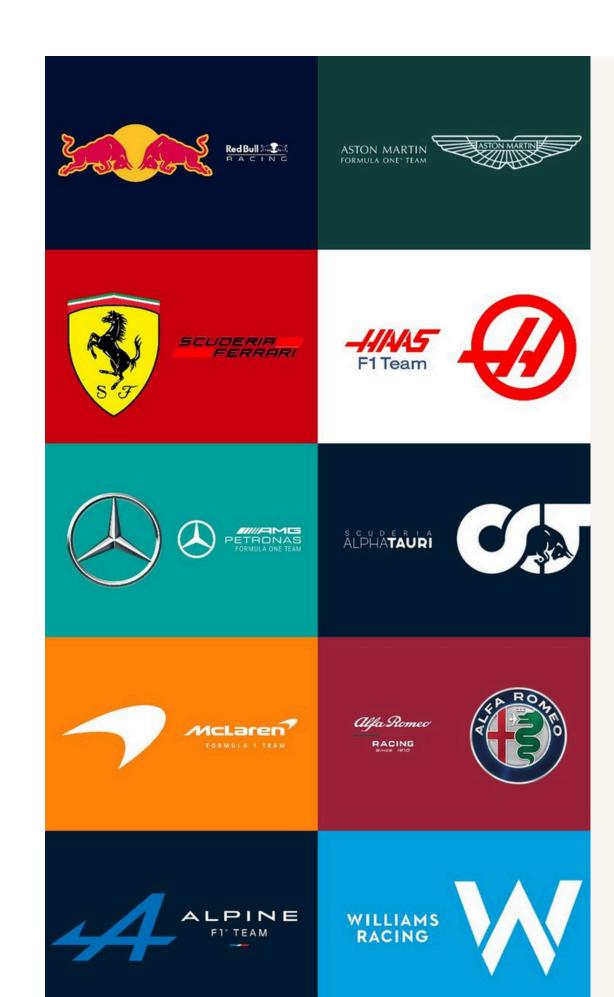


Table of contents

DATA VISUALIZATION 04

PIT STOP CHAMPIONSHIP 06

MODELS 10

CONCLUSION 13

Data Visualization

• Several dataframes: lap times, pit stops, circuits, drivers, races, status, driver standings, results, and constructors linked together through driver IDs and race IDs.

	raceld	driverId	stop	lap	time	duration	milliseconds
0	841	153	1	1	17:05:23	26.898	26898
1	841	30	1	1	17:05:52	25.021	25021
2	841	17	1	11	17:20:48	23.426	23426
3	841	4	1	12	17:22:34	23.251	23251
4	841	13	1	13	17:24:10	23.842	23842

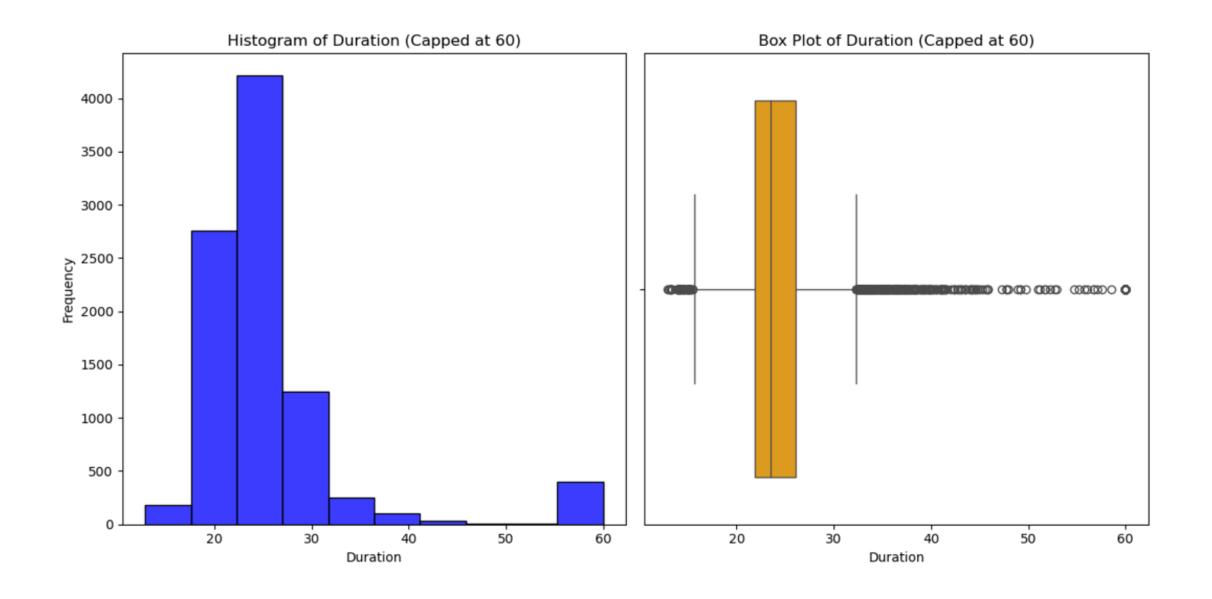
Original pit stop dataframe

- "Duration" converted to **float**
- No NaN
- 9203 observations

driverRef	year	name	duration	stop	
alguersuari	2011	Australian Grand Prix	26.898	1	0
webber	2011	Australian Grand Prix	23.426	1	1
alonso	2011	Australian Grand Prix	23.251	1	2
massa	2011	Australian Grand Prix	23.842	1	3
vettel	2011	Australian Grand Prix	22.603	1	4

Transformed pit stop dataframe

Data Visualization

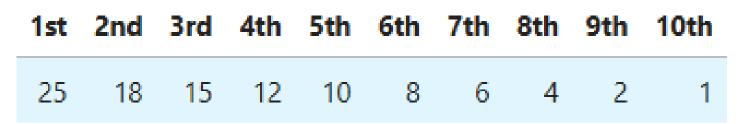


- mean = 82.347, median = 23.529
- many outliers

PIT STOP CHAMPIONSHIP

Hypothesis

- If a driver doesn't finish the race or finishes with a delay of more than 3 laps, he is considered as
 DNF
- We consider only the seasons between 2011 and 2023
- We rank the drivers by their average duration in the pit lane for each race
- We award the same number of points as the official championship





Results

	points	pit_points	rank_std	pit_rank
driverRef				
hamilton	384.0	65	1	13
rosberg	317.0	86	2	10
ricciardo	238.0	219	3	1
bottas	186.0	72	4	11
vettel	167.0	185	5	4
alonso	161.0	208	6	2
massa	134.0	113	7	7
button	126.0	161	8	5
hulkenberg	96.0	99	9	9
perez	59.0	58	10	14
kevin_magnussen	55.0	200	11	3
raikkonen	55.0	118	12	6
vergne	22.0	20	13	19
kvyat	8.0	31	15	16
grosjean	8.0	106	14	8
jules_bianchi	2.0	2	17	22
maldonado	2.0	67	16	12
stevens	0.0	0	22	23
sutil	0.0	12	23	20
gutierrez	0.0	20	24	18

	points	pit_points	rank_std	pit_rank
driverRef				
hamilton	381.0	271	1	1
rosberg	322.0	209	2	3
vettel	278.0	239	3	2
raikkonen	150.0	155	4	4
bottas	136.0	96	5	8
massa	121.0	90	6	9
kvyat	95.0	123	7	5
ricciardo	92.0	102	8	7
perez	78.0	108	9	6
hulkenberg	58.0	86	10	10
grosjean	51.0	47	11	16
max_verstappen	49.0	73	12	12
nasr	27.0	34	13	17
maldonado	27.0	69	14	13
sainz	18.0	16	15	18
button	16.0	80	16	11
alonso	11.0	66	17	14
ericsson	9.0	49	18	15
merhi	0.0	6	19	19
rossi	0.0	0	21	20

Results

	points	pit_points	rank_std	pit_rank
constructorRef				
mercedes	701.0	151	1	7
red_bull	405.0	404	2	1
williams	320.0	185	3	4
ferrari	216.0	326	4	3
mclaren	181.0	361	5	2
force_india	155.0	157	6	6
toro_rosso	30.0	51	7	8
lotus_f1	10.0	173	8	5
marussia	2.0	31	9	11
caterham	0.0	48	10	9
sauber	0.0	32	11	10

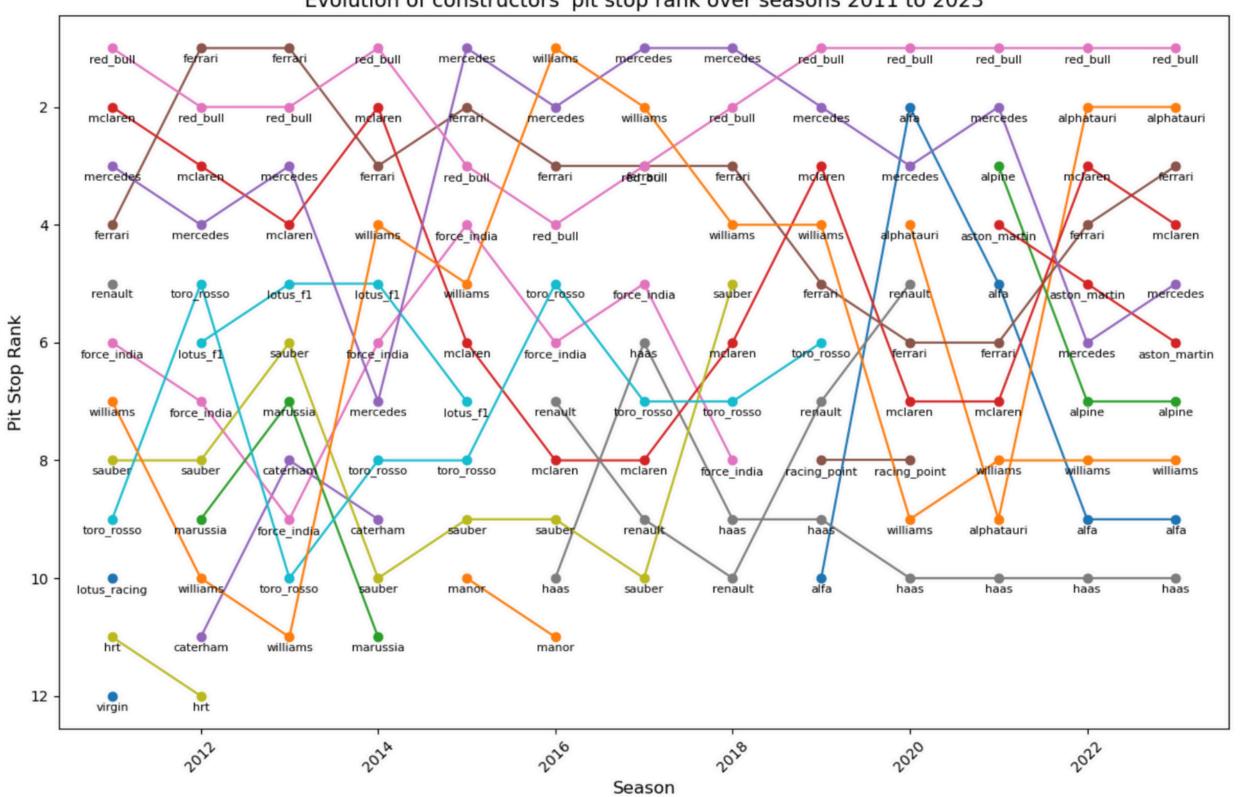
Constructors' Ranking 2014

points	pit_points	rank_std	pit_rank
703.0	480	1	1
428.0	394	2	2
257.0	186	3	5
187.0	225	4	3
136.0	194	5	4
78.0	116	6	7
67.0	89	7	8
36.0	83	8	9
27.0	146	9	6
0.0	6	10	10
	703.0 428.0 257.0 187.0 136.0 78.0 67.0 36.0	703.0 480 428.0 394 257.0 186 187.0 225 136.0 194 78.0 116 67.0 89 36.0 83 27.0 146	428.0 394 2 257.0 186 3 187.0 225 4 136.0 194 5 78.0 116 6 67.0 89 7 36.0 83 8 27.0 146 9

Constructors' Ranking 2015

Results

Evolution of constructors' pit stop rank over seasons 2011 to 2023



Correlation between Constructors' Pit Stop Championship and official constructors' championship:

- Pearson correlation: 0.7554 (points)
- Spearman correlation: 0.7380 (ranks)
- Kendall-Tau correlation: 0.5980 (ranks)

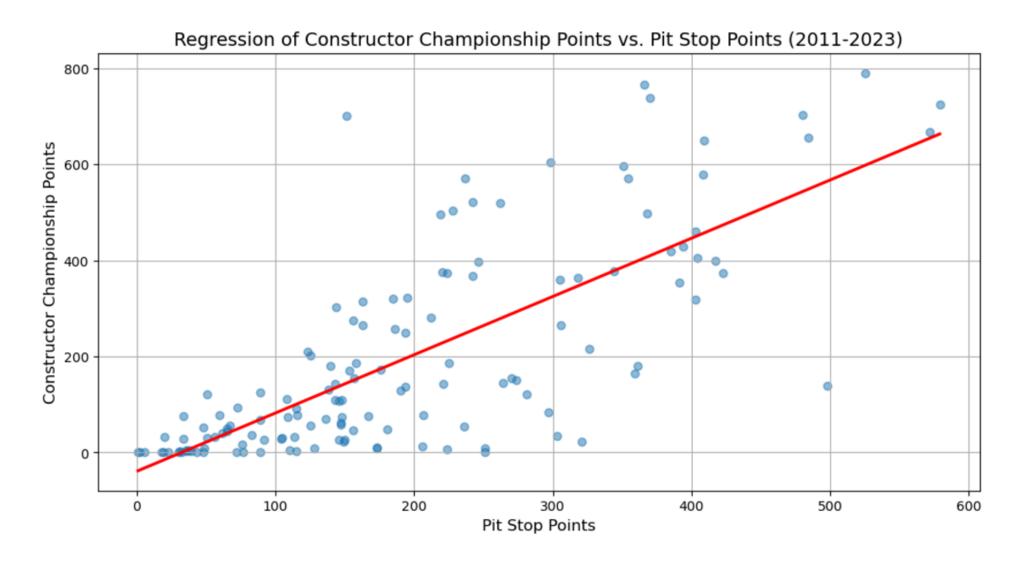
Correlation between Drivers' Pit Stop Championship and official drivers' championship:

- Pearson correlation: 0.7443 (points)
- Spearman correlation: 0.7584 (ranks)
- Kendall-Tau correlation: 0.5842 (ranks)

OLS REGRESSION FOR CONSTRUCTORS

OLS Regression Results

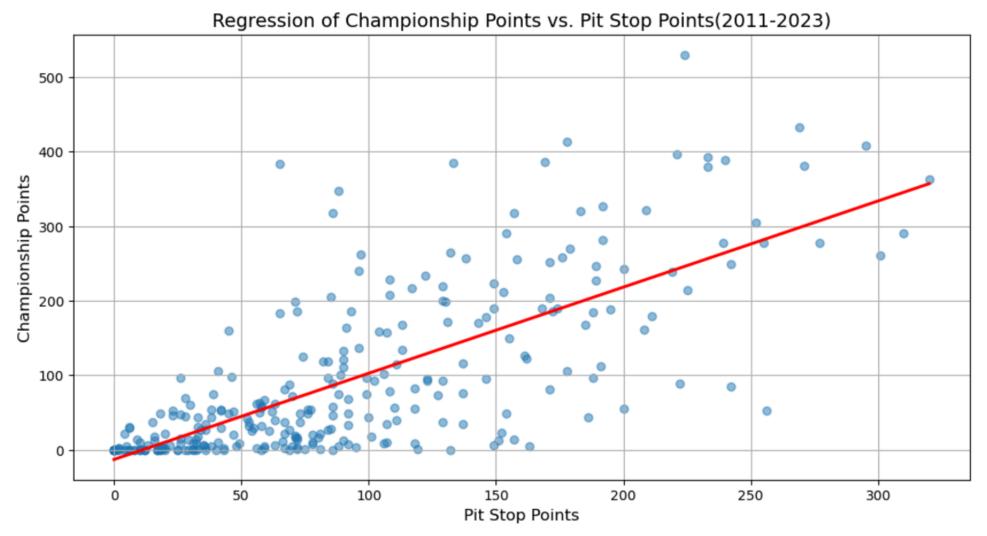
======================================								
Dep. Variable	e:	ро	ints	R-sq	uared:		0.571	
Model:		-	OLS	Adj.	R-squared:		0.567	
Method:		Least Squares		F-st	atistic:		179.4	
Date:		Tue, 26 Nov 2024		Prob	(F-statistic)	:	1.48e-26	
Time:		17:3	5:05	Log-	Likelihood:		-870.99	
No. Observati	ons:		137	AIC:			1746.	
Df Residuals:			135	BIC:			1752.	
Df Model:			1					
Covariance Type:		nonro	bust					
=========			=====			=======	=======	
	coef	std err		t	P> t	[0.025	0.975]	
const	-39.1702	21.163	-1	1.851	0.066	-81.024	2.683	
pit_points	1.2129	0.091	13	3.395	0.000	1.034	1.392	
Omnibus:		 13	.758	Durb	in-Watson:		1.857	
Prob(Omnibus):		0.001		Jarque-Bera (JB):			24.504	
Skew:		0	.448	Prob	(JB):		4.78e-06	
Kurtosis:		4	.868	Cond	. No.		412.	



OLS REGRESSION FOR DRIVERS

OLS Regression Results

Dep. Variable: points			oints	R-squ	R-squared:			
Model:		OLS	Adj.	R-squared:		0.552		
Method:		Least Squares		F-sta	atistic:		362.7	
Date: Tue,		Tue, 26 Nov	26 Nov 2024		(F-statistic)	3.94e-53		
Time:		17:3	34:02	Log-l	ikelihood:		-1682.5	
No. Observat	ions:		294	AIC:			3369.	
Df Residuals		292	BIC:			3376.		
Df Model:		1						
Covariance T	ype:	nonro	bust					
========								
	coef			t	P> t	[0.025	0.975]	
const	-13.1937				0.059	-26.883	0.496	
pit_points	1.1565	0.061	19	9.046	0.000	1.037	1.276	
Omnibus:		49	 9.363	Durbi	in-Watson:	======	1.447	
Prob(Omnibus):		6	0.000	Jarqu	ue-Bera (JB):		120.796	
Skew:		6	797	Prob((JB):		5.88e-27	
Kurtosis:		9	5.705	Cond.	. No.		184.	



IMPROVEMENT AND EXTENSIONS

- Adding more variables in the OLS regression (e.g. age, years in F1, wet or dry race, ...)
- Use other types of regression like quadratic regression, spline regression,...
- Use another variable to represent the pit stop efficiency
- Stopping time instead of the time spent in the pit lane to isolate the pit crew technicians efficiency
- Analyze relationships across seasons to identify trends over time
- Investigate causal relationships

Conclusion

- Top teams consistently perform better
- High correlation and R-squared between our Pit Stop Championship and the official championships
- Seasonal variability and model improvement

