

Predicting the 2021 Fantasy Football Draft ARIMA

May 4, 2021

- 1 Who should be the first pick drafted in the 2021 NFL Fantasy Draft? This project aims to predict players total points for next season to provide an objective ranking.

```
[1]: # Import Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns; sns.set()
import sklearn
```

```
[2]: # df_2020 = pd.read_csv('2020NFLFantasyStats.csv', header = 1)
# df_2020['Player'] = df_2020['Player'].str.split('\\').str[0].str.
    ↳replace('[+,*]', '')
# df_2019 = pd.read_csv('2019NFLFantasyStats.csv', header = 1)
# df_2019['Player'] = df_2019['Player'].str.split('\\').str[0].str.
    ↳replace('[+,*]', '')
```

```
[3]: # Load in the Data
list_of_names = _
    ↳['2020NFLFantasyStats', '2019NFLFantasyStats', '2018NFLFantasyStats', '2017NFLFantasyStats', '2016NFLFantasyStats']
df_list = []

for i in range(len(list_of_names)):
    temp_df = pd.read_csv( list_of_names[i] + '.csv', header = 1)
    temp_df['Player'] = temp_df['Player'].str.split('\\').str[0].str.
    ↳replace('[+,*]', '').str.strip()
    df_list.append(temp_df)
```

```
[4]: # Add the year column to each dataset
for dataset in df_list:
    df_2020 = df_list[0]
    df_2020['Year'] = 2020
    df_2019 = df_list[1]
    df_2019['Year'] = 2019
    df_2018 = df_list[2]
```

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df_2018['Year'] = 2018
df_2017 = df_list[3]
df_2017['Year'] = 2017
df_2016 = df_list[4]
df_2016['Year'] = 2016
df_2015 = df_list[5]
df_2015['Year'] = 2015
df_2014 = df_list[6]
df_2014['Year'] = 2014
df_2013 = df_list[7]
df_2013['Year'] = 2013
df_2012 = df_list[8]
df_2012['Year'] = 2012
df_2011 = df_list[9]
df_2011['Year'] = 2011
display(dataset)

```

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	2PM	\
0	1	Derrick Henry	TEN	RB	26	16	16	0	0	0	...	1.0	
1	2	Alvin Kamara	NOR	RB	25	15	10	0	0	0	...	NaN	
2	3	Dalvin Cook	MIN	RB	25	14	14	0	0	0	...	3.0	
3	4	Travis Kelce	KAN	TE	31	15	15	1	2	4	...	1.0	
4	5	Davante Adams	GNB	WR	28	14	14	0	0	0	...	NaN	
..	
621	622	Andre Roberts	BUF	WR	32	15	0	0	0	0	...	NaN	
622	623	Nate Sudfeld	PHI	QB	27	1	0	5	12	32	...	NaN	
623	624	Nsimba Webster	LAR	NaN	24	16	0	0	0	0	...	NaN	
624	625	Dede Westbrook	JAX	WR	27	2	0	0	0	0	...	NaN	
625	626	Kendall Hinton	DEN	WR	23	1	0	1	9	13	...	NaN	

	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	314.0	333.1	341.1	323.6	184.0	1	1.0	2020
1	NaN	295.0	377.8	383.8	336.3	165.0	2	2.0	2020
2	NaN	294.0	337.8	346.8	315.8	164.0	3	3.0	2020
3	NaN	208.0	312.8	316.8	260.3	117.0	1	4.0	2020
4	NaN	243.0	358.4	362.4	300.9	117.0	1	5.0	2020
..
621	NaN	-1.0	3.1	5.1	1.1	NaN	245	NaN	2020
622	NaN	-2.0	-1.5	0.5	-0.5	NaN	81	NaN	2020
623	NaN	-2.0	-2.0	-1.0	-2.0	NaN	248	NaN	2020
624	NaN	-2.0	-0.6	0.4	-1.1	NaN	247	NaN	2020
625	NaN	-3.0	-2.8	-0.8	-0.8	NaN	249	NaN	2020

[626 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Christian McCaffrey	CAR	RB	23	16	16	0	2	0	...	

1	2	Lamar Jackson	BAL	QB	22	15	15	265	401	3127	...
2	3	Derrick Henry	TEN	RB	25	15	15	0	0	0	...
3	4	Aaron Jones	GNB	RB	25	16	16	0	0	0	...
4	5	Ezekiel Elliott	DAL	RB	24	16	16	0	0	0	...
..
614	615	Ray-Ray McCloud	CAR	NaN	23	6	0	0	0	0	...
615	616	Darrius Shepherd	GNB	WR	24	6	0	0	0	0	...
616	617	Jarrett Stidham	NWE	QB	23	3	0	2	4	14	...
617	618	Michael Walker	JAX	WR	23	7	0	0	0	0	...
618	619	Corey Clement	PHI	NaN	25	4	0	0	0	0	...

	2PM	2PP	FantPt	PPR	DKPt	FDPT	VBD	PosRank	OvRank	Year
0	1.0	NaN	355.0	471.2	477.2	413.2	215.0	1	1.0	2019
1	NaN	NaN	416.0	415.7	429.7	421.7	152.0	1	2.0	2019
2	NaN	NaN	277.0	294.6	303.6	285.6	136.0	2	3.0	2019
3	NaN	NaN	266.0	314.8	322.8	290.3	125.0	3	4.0	2019
4	NaN	NaN	258.0	311.7	319.7	284.7	117.0	4	5.0	2019
..
614	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	244	NaN	2019
615	NaN	NaN	-2.0	-0.9	0.1	-1.4	NaN	242	NaN	2019
616	NaN	NaN	-2.0	-1.6	-0.6	-0.6	NaN	73	NaN	2019
617	NaN	NaN	-3.0	-0.5	1.5	-1.5	NaN	246	NaN	2019
618	NaN	NaN	-4.0	-4.0	-2.0	-4.0	NaN	165	NaN	2019

[619 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Todd Gurley	LAR	RB	24	14	14	0	0	0	...	
1	2	Saquon Barkley	NYG	RB	21	16	16	0	0	0	...	
2	3	Christian McCaffrey	CAR	RB	22	16	16	1	1	50	...	
3	4	Alvin Kamara	NOR	RB	23	15	13	0	0	0	...	
4	5	Patrick Mahomes	KAN	QB	23	16	16	383	580	5097	...	
..	
603	604	Kaelin Clay	NYG	NaN	26	2	0	0	0	0	...	
604	605	JJ Jones	2TM	WR	26	4	0	0	0	0	...	
605	606	Kyle Lauletta	NYG	QB	23	2	0	0	5	0	...	
606	607	Riley McCarron	NWE	NaN	25	1	0	0	0	0	...	
607	608	Jojo Natson	LAR	NaN	24	13	0	0	0	0	...	

	2PM	2PP	FantPt	PPR	DKPt	FDPT	VBD	PosRank	OvRank	Year
0	3.0	NaN	313.0	372.1	379.1	342.6	178.0	1	1.0	2018
1	1.0	NaN	295.0	385.8	391.8	340.3	159.0	2	2.0	2018
2	NaN	NaN	279.0	385.5	392.5	332.0	143.0	3	3.0	2018
3	3.0	NaN	273.0	354.2	360.2	313.7	138.0	4	4.0	2018
4	1.0	NaN	417.0	417.1	437.1	429.1	134.0	1	5.0	2018
..
603	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	242	NaN	2018

604	NaN	NaN	-2.0	-0.7	0.3	-1.2	NaN	239	NaN	2018
605	NaN	NaN	-2.0	-2.2	-1.2	-1.2	NaN	75	NaN	2018
606	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	241	NaN	2018
607	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	240	NaN	2018

[608 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Todd Gurley	LAR	RB	23	15	15	0	0	0	...	
1	2	Le'Veon Bell	PIT	RB	25	15	15	0	0	0	...	
2	3	Kareem Hunt	KAN	RB	22	16	16	0	0	0	...	
3	4	Alvin Kamara	NOR	RB	22	16	3	0	0	0	...	
4	5	Melvin Gordon	LAC	RB	24	16	16	0	0	0	...	
..	
584	585	John Ross	CIN	WR	22	3	1	0	0	0	...	
585	586	Jake Rudock	DET	QB	24	3	0	3	5	24	...	
586	587	Tyler Bray	KAN	QB	26	1	0	0	1	0	...	
587	588	Teddy Bridgewater	MIN	QB	25	1	0	0	2	0	...	
588	589	Devante Mays	GNB	RB	23	8	0	0	0	0	...	

	2PM	2PP	FantPt	PPR	DKPt	FDPT	VBD	PosRank	OvRank	Year
0	NaN	NaN	319.0	383.3	391.3	351.3	192.0	1	1.0	2017
1	NaN	NaN	257.0	341.6	349.6	299.1	130.0	2	2.0	2017
2	NaN	NaN	242.0	295.2	302.2	268.7	115.0	3	3.0	2017
3	1.0	NaN	239.0	320.4	327.4	279.9	112.0	4	4.0	2017
4	NaN	NaN	230.0	288.1	294.1	259.1	103.0	5	5.0	2017
..
584	NaN	NaN	-1.0	-0.8	0.2	-0.8	NaN	224	NaN	2017
585	NaN	NaN	-1.0	-1.0	NaN	NaN	NaN	72	NaN	2017
586	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	74	NaN	2017
587	NaN	NaN	-2.0	-2.3	-1.3	-1.3	NaN	75	NaN	2017
588	NaN	NaN	-2.0	1.1	2.1	-0.4	NaN	160	NaN	2017

[589 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	2PM	\
0	1	David Johnson	ARI	RB	25	16	16	0	0	0	...	1.0	
1	2	Ezekiel Elliott	DAL	RB	21	15	15	0	0	0	...	NaN	
2	3	Aaron Rodgers	GNB	QB	33	16	16	401	610	4428	...	1.0	
3	4	LeSean McCoy	BUF	RB	28	15	15	0	0	0	...	1.0	
4	5	Le'Veon Bell	PIT	RB	24	12	12	0	1	0	...	1.0	
..	
569	570	Tyler Ervin	HOU	RB	23	12	0	0	0	0	...	NaN	
570	571	Brett Hundley	GNB	QB	23	4	0	2	10	17	...	NaN	
571	572	Keshawn Martin	SFO	NaN	26	1	0	0	0	0	...	NaN	
572	573	Kalif Raymond	DEN	NaN	22	4	0	0	0	0	...	NaN	
573	574	Matthew Slater	NWE	WR	31	13	0	0	0	0	...	NaN	

	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	328.0	407.8	416.8	367.8	191.0	1	1.0	2016
1	NaN	293.0	325.4	332.4	309.4	156.0	2	2.0	2016
2	1.0	380.0	380.0	397.0	387.0	119.0	1	3.0	2016
3	NaN	248.0	298.3	304.3	273.3	111.0	3	4.0	2016
4	NaN	242.0	317.4	324.4	279.9	105.0	4	5.0	2016
..
569	NaN	-2.0	1.1	3.1	-0.4	NaN	160	NaN	2016
570	NaN	-2.0	-1.5	-0.5	-0.5	NaN	70	NaN	2016
571	NaN	-2.0	-2.0	-1.0	-2.0	NaN	213	NaN	2016
572	NaN	-2.0	-2.0	-1.0	-2.0	NaN	214	NaN	2016
573	NaN	-2.0	-1.5	-0.5	-1.5	NaN	212	NaN	2016

[574 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	2PM	\
0	1	Antonio Brown	PIT	WR	27	16	16	0	0	0	...	2.0	
1	2	Devonta Freeman	ATL	RB	23	15	13	0	0	0	...	NaN	
2	3	Julio Jones	ATL	WR	26	16	16	0	0	0	...	NaN	
3	4	Adrian Peterson	MIN	RB	30	16	16	0	0	0	...	NaN	
4	5	Cam Newton	CAR	QB	26	16	16	296	495	3837	...	NaN	
..	
569	570	Frankie Hammond	KAN	NaN	25	9	0	0	0	0	...	NaN	
570	571	Raheem Mostert	3TM	NaN	23	11	0	0	0	0	...	NaN	
571	572	Isaiah Pead	STL	RB	26	2	0	0	0	0	...	NaN	
572	573	Sean Renfree	ATL	QB	25	2	0	3	7	11	...	NaN	
573	574	Glenn Winston	CLE	RB	26	3	0	0	0	0	...	NaN	

	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	252.0	388.2	393.2	320.2	125.0	1	1.0	2015
1	NaN	243.0	316.4	324.4	279.9	123.0	1	2.0	2015
2	NaN	239.0	375.1	379.1	307.1	111.0	2	3.0	2015
3	NaN	231.0	260.7	269.7	245.7	110.0	2	4.0	2015
4	NaN	389.0	389.1	409.1	399.1	105.0	1	5.0	2015
..
569	NaN	-2.0	-2.0	-1.0	-2.0	NaN	206	NaN	2015
570	NaN	-2.0	-2.0	-1.0	-2.0	NaN	164	NaN	2015
571	NaN	-2.0	-1.7	-0.7	-1.7	NaN	163	NaN	2015
572	NaN	-2.0	-2.0	-1.0	-1.0	NaN	73	NaN	2015
573	NaN	-3.0	-2.8	-1.8	-2.8	NaN	165	NaN	2015

[574 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	2PM	\
0	1	DeMarco Murray	DAL	RB	26	16	16.0	0	0	0	...	NaN	
1	2	Le'Veon Bell	PIT	RB	22	16	16.0	0	0	0	...	NaN	

2	3	Marshawn Lynch	SEA	RB	28	16	14.0	0	0	0	...	NaN
3	4	Antonio Brown	PIT	WR	26	16	16.0	2	2	20	...	1.0
4	5	Matt Forte	CHI	RB	29	16	16.0	0	0	0	...	2.0
..
756	757	Isaiah Burse	DEN	NaN	23	12	0.0	0	0	0	...	NaN
757	758	Matt Flynn	GNB	QB	29	7	0.0	8	16	66	...	NaN
758	759	Tauren Poole	3TM	RB	25	3	0.0	0	0	0	...	NaN
759	760	Walt Powell	NYJ	NaN	23	4	0.0	0	0	0	...	NaN
760	761	Matt Schaub	OAK	QB	33	11	0.0	5	10	57	...	NaN

	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	294.0	351.1	362.1	322.6	173.0	1	1.0	2014
1	NaN	288.0	370.5	376.5	329.0	167.0	2	2.0	2014
2	NaN	265.0	302.3	310.3	283.8	144.0	3	3.0	2014
3	NaN	258.0	386.9	391.9	322.4	136.0	1	4.0	2014
4	NaN	245.0	346.6	354.6	295.6	124.0	4	5.0	2014
..
756	NaN	-2.0	-2.0	-1.0	-2.0	NaN	289	NaN	2014
757	NaN	-2.0	-2.4	-0.4	-1.4	NaN	103	NaN	2014
758	NaN	-2.0	-2.0	-1.0	-2.0	NaN	199	NaN	2014
759	NaN	-2.0	-2.0	-1.0	-2.0	NaN	288	NaN	2014
760	NaN	-4.0	-3.7	-0.7	-1.7	NaN	104	NaN	2014

[761 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Jamaal Charles	KAN	RB	27	15	15	0	0	0	...	
1	2	Peyton Manning	DEN	QB	37	16	16	450	659	5477	...	
2	3	LeSean McCoy	PHI	RB	25	16	16	0	0	0	...	
3	4	Matt Forte	CHI	RB	28	16	16	0	0	0	...	
4	5	Jimmy Graham	NOR	TE	27	16	12	0	0	0	...	
..	
575	576	Charlie Whitehurst	SDG	QB	31	2	0	0	0	0	...	
576	577	Tyler Clutts	2TM	RB	29	8	0	0	0	0	...	
577	578	Matt Moore	MIA	QB	29	1	0	2	6	53	...	
578	579	Curtis Painter	NYG	QB	28	3	0	8	16	57	...	
579	580	Devon Wylie	TEN	NaN	25	2	0	0	0	0	...	

	2PM	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	NaN	308.0	378.0	386.0	343.0	182.0	1	1.0	2013
1	NaN	1.0	412.0	412.0	431.0	422.0	153.0	1	2.0	2013
2	NaN	NaN	279.0	330.6	337.6	304.6	152.0	2	3.0	2013
3	1.0	NaN	263.0	337.3	345.3	300.3	137.0	3	4.0	2013
4	NaN	NaN	218.0	303.5	306.5	260.5	124.0	1	5.0	2013
..
575	NaN	NaN	-1.0	-0.5	-0.5	-0.5	NaN	67	NaN	2013
576	NaN	NaN	-2.0	-0.6	0.4	-1.1	NaN	170	NaN	2013

577	NaN	NaN	-2.0	-1.9	0.1	0.1	NaN	68	NaN	2013
578	NaN	NaN	-2.0	-1.9	0.1	0.1	NaN	69	NaN	2013
579	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	214	NaN	2013

[580 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Adrian Peterson	MIN	RB	27	16.0	16.0	0.0	0.0	0.0	...	
1	2	Doug Martin	TAM	RB	23	16.0	16.0	0.0	0.0	0.0	...	
2	3	Arian Foster	HOU	RB	26	16.0	16.0	0.0	0.0	0.0	...	
3	4	Marshawn Lynch	SEA	RB	26	16.0	15.0	0.0	0.0	0.0	...	
4	5	Alfred Morris	WAS	RB	24	16.0	16.0	0.0	0.0	0.0	...	
..	
682	683	Brock Osweiler	DEN	QB	22	5.0	0.0	2.0	4.0	12.0	...	
683	684	Graham Harrell	GNB	QB	27	4.0	0.0	2.0	4.0	20.0	...	
684	685	Josh Johnson	CLE	NaN	26	1.0	0.0	0.0	0.0	0.0	...	
685	686	Ryan Mallett	NWE	QB	24	4.0	0.0	1.0	4.0	17.0	...	
686	687	Joe Adams	CAR	WR	23	9.0	0.0	0.0	0.0	0.0	...	

	2PM	2PP	FantPt	PPR	DKPt	FDPT	VBD	PosRank	OvRank	Year
0	1.0	NaN	307.0	347.4	355.4	327.4	191.0	1	1.0	2012
1	NaN	NaN	263.0	311.6	318.6	287.1	146.0	2	2.0	2012
2	NaN	NaN	262.0	302.1	310.1	282.1	145.0	3	3.0	2012
3	NaN	NaN	247.0	269.6	277.6	258.1	130.0	4	4.0	2012
4	NaN	NaN	241.0	252.0	258.0	246.5	124.0	5	5.0	2012
..
682	NaN	NaN	-1.0	-0.8	-0.8	-0.8	NaN	85	NaN	2012
683	NaN	NaN	-2.0	-1.5	-0.5	-1.5	NaN	86	NaN	2012
684	NaN	NaN	-2.0	-2.0	-1.0	-2.0	NaN	87	NaN	2012
685	NaN	NaN	-2.0	-2.2	-1.2	-1.2	NaN	88	NaN	2012
686	NaN	NaN	-4.0	-3.0	NaN	-3.5	NaN	254	NaN	2012

[687 rows x 34 columns]

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
0	1	Aaron Rodgers	GNB	QB	28	15	15	343.0	502.0	4643.0	...	
1	2	Drew Brees	NOR	QB	32	16	16	468.0	657.0	5476.0	...	
2	3	Ray Rice	BAL	RB	24	16	16	1.0	1.0	1.0	...	
3	4	Calvin Johnson	DET	WR	26	16	16	0.0	0.0	0.0	...	
4	5	Cam Newton	CAR	QB	22	16	16	310.0	517.0	4051.0	...	
..	
578	579	Kris Wilson	BAL	NaN	30	15	0	0.0	0.0	0.0	...	
579	580	Will Yeatman	MIA	NaN	23	3	2	0.0	0.0	0.0	...	
580	581	Joe Zelenka	ATL	NaN	35	16	0	0.0	0.0	0.0	...	
581	582	Billy Volek	SDG	QB	35	2	0	0.0	0.0	0.0	...	
582	583	David Reed	BAL	WR	24	11	0	0.0	0.0	0.0	...	

	2PM	2PP	FantPt	PPR	DKPt	FDPt	VBD	PosRank	OvRank	Year
0	NaN	NaN	397.0	397.4	409.4	403.4	175.0	1	1.0	2011
1	NaN	3.0	394.0	393.6	411.6	407.6	171.0	2	2.0	2011
2	NaN	NaN	297.0	372.8	380.8	334.8	149.0	1	3.0	2011
3	NaN	NaN	263.0	359.2	363.2	311.2	148.0	1	4.0	2011
4	NaN	2.0	369.0	370.3	395.3	386.8	147.0	3	5.0	2011
...
578	NaN	NaN	NaN	NaN	NaN	NaN	NaN	108	NaN	2011
579	NaN	NaN	NaN	NaN	NaN	NaN	NaN	104	NaN	2011
580	NaN	NaN	NaN	NaN	NaN	NaN	NaN	115	NaN	2011
581	NaN	NaN	-1.0	-0.5	-0.5	-0.5	NaN	77	NaN	2011
582	NaN	NaN	-2.0	-2.4	-0.4	-2.4	NaN	213	NaN	2011

[583 rows x 34 columns]

```
[5]: # Concat 2011-19 dataframes for the training data set
frames = [df_2019,
          df_2018,df_2017,df_2016,df_2015,df_2014,df_2013,df_2012,df_2011]
train_df_2011_2019 = pd.concat(frames,ignore_index = True)
# Test data set will be the 2020 season
test_df = df_2020
```

```
[6]: # Create Dataframes for each position
RB_df = train_df_2011_2019[train_df_2011_2019['FantPos'] == 'RB']
TE_df = train_df_2011_2019[train_df_2011_2019['FantPos'] == 'TE']
WR_df = train_df_2011_2019[train_df_2011_2019['FantPos'] == 'WR']
QB_df = train_df_2011_2019[train_df_2011_2019['FantPos'] == 'QB']
```

```
[7]: # Enter a string of the player you want stats for
def player(player):
    player_df = train_df_2011_2019[train_df_2011_2019['Player'] == player]
    return player_df
```

```
[8]: # Tom Brady Example
Tom_Brady_df = player('Tom Brady')
Tom_Brady_df
```

```
[8]:
```

	Rk	Player	Tm	FantPos	Age	G	GS	Cmp	Att	Yds	...	\
78	79	Tom Brady	NWE	QB	42	16.0	16.0	373.0	613.0	4057.0	...	
698	80	Tom Brady	NWE	QB	41	16.0	16.0	375.0	570.0	4355.0	...	
1254	28	Tom Brady	NWE	QB	40	16.0	16.0	385.0	581.0	4577.0	...	
1895	80	Tom Brady	NWE	QB	39	12.0	12.0	291.0	432.0	3554.0	...	
2410	21	Tom Brady	NWE	QB	38	16.0	16.0	402.0	624.0	4770.0	...	
3025	62	Tom Brady	NWE	QB	37	16.0	16.0	373.0	582.0	4109.0	...	
3805	81	Tom Brady	NWE	QB	36	16.0	16.0	380.0	628.0	4343.0	...	

4316	12	Tom Brady	NWE	QB	35	16.0	16.0	401.0	637.0	4827.0	...
4998	7	Tom Brady	NWE	QB	34	16.0	16.0	401.0	611.0	5235.0	...

	2PM	2PP	FantPt	PPR	DKPt	FDPT	VBD	PosRank	OvRank	Year
78	NaN	1.0	264.0	263.7	275.7	271.7	NaN	12	NaN	2019
698	NaN	NaN	280.0	281.3	297.3	291.8	NaN	14	NaN	2018
1254	NaN	2.0	296.0	295.9	309.9	303.9	41.0	3	28.0	2017
1895	NaN	1.0	259.0	258.6	263.6	260.6	NaN	15	NaN	2016
2410	NaN	NaN	344.0	344.7	356.7	351.2	59.0	2	21.0	2015
3025	NaN	NaN	278.0	278.1	293.1	287.1	12.0	9	62.0	2014
3805	NaN	2.0	252.0	251.5	268.5	262.5	NaN	14	NaN	2013
4316	NaN	NaN	340.0	340.3	351.3	348.3	90.0	3	12.0	2012
4998	NaN	NaN	366.0	366.3	386.3	378.3	144.0	4	7.0	2011

[9 rows x 34 columns]

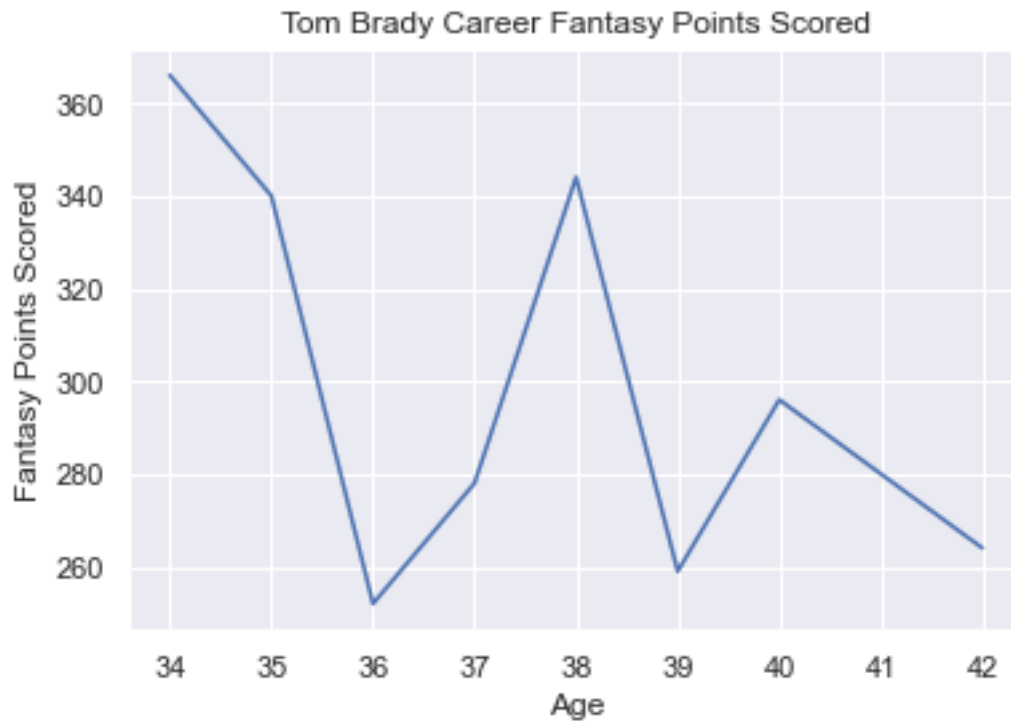
```
[9]: # Plot Function
def plot_f(player_name):

    df = player(player_name)

    Age = df['Age']
    Points = df['FantPt']

    plt.plot(Age, Points)
    plt.xlabel('Age')
    plt.ylabel('Fantasy Points Scored')
    plt.title(player_name + ' Career Fantasy Points Scored')
```

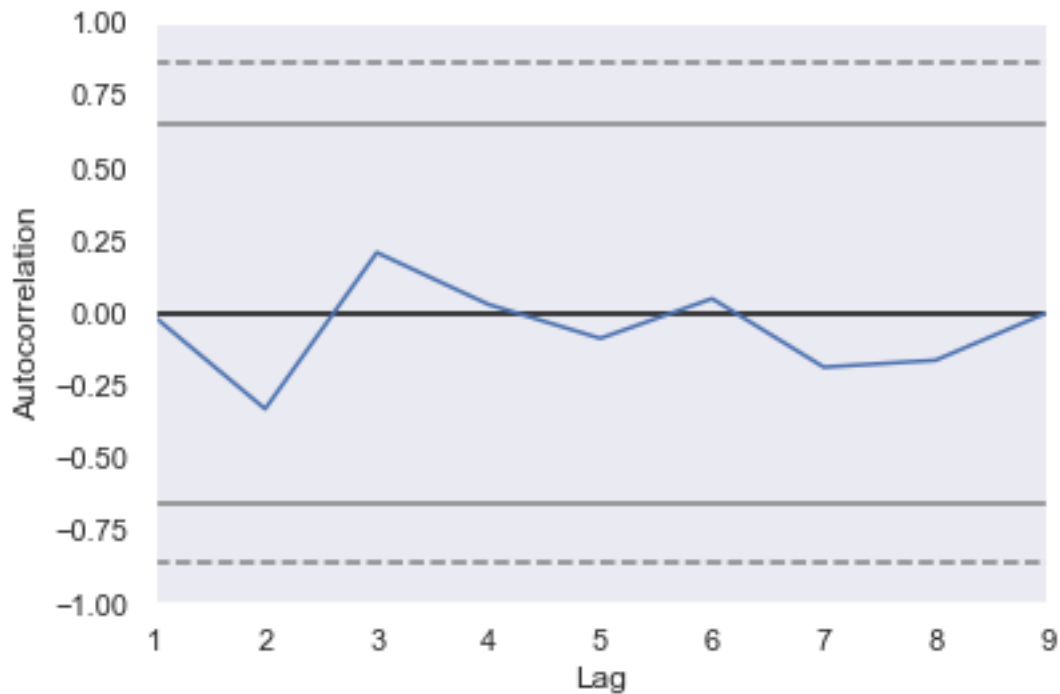
```
[10]: plot_f('Tom Brady')
```



```
[11]: from statsmodels.tsa.arima.model import ARIMA
      from sklearn.metrics import mean_squared_error
      from pandas.plotting import autocorrelation_plot

      # ACF Plot Function
      def acf_plot_f(player_name):
          autocorrelation_plot(player(player_name).FantPt)
          plt.show()
```

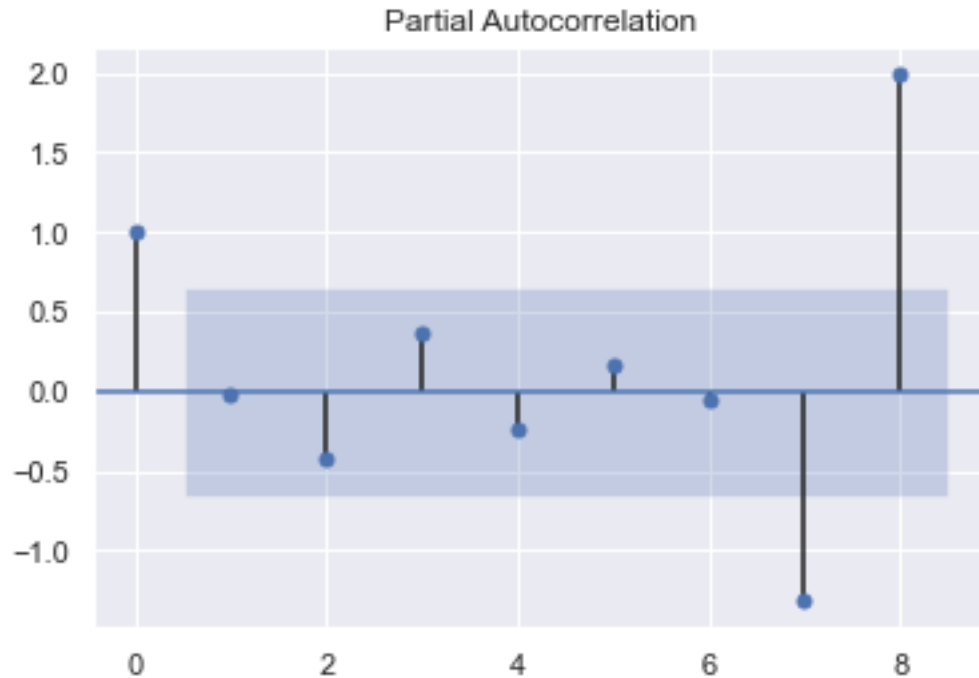
```
[12]: acf_plot_f('Tom Brady')
```



```
[13]: from statsmodels.graphics.tsaplots import plot_pacf
      # Partial Autocorrelation Function
      def pacf_plot_f(player_name):
          plot_pacf(player(player_name).FantPt)
          plt.show()
```

```
[14]: pacf_plot_f('Tom Brady')
```

```
/Users/nickseah/Desktop/PythonAddOns/anaconda3/lib/python3.8/site-
packages/statsmodels/regression/linear_model.py:1406: RuntimeWarning: invalid
value encountered in sqrt
    return rho, np.sqrt(sigmatq)
```



```
[15]: # ARIMA Model
# import warnings

# # Function to forecast number of fantasy points in a standard fantasy league

def forecast_f(player_name):
    qb_model = ARIMA(player(player_name).FantPt, order = (2,1,0))
    model_fit = qb_model.fit()
    forecast = model_fit.forecast(steps = 1)

    return print(forecast)
```

```
[16]: # # ARIMA Model
# import warnings

# Function to forecast number of fantasy points in a standard fantasy league

# def rb_forecast_f(player_name):
#     rb_model = ARIMA(player(player_name).FantPt, order = (2,1,0))
#     model_fit = rb_model.fit()
#     forecast = model_fit.forecast(steps = 1)
```

```
#     return print(forecast)
```

```
[17]: # ARIMA Model
# import warnings

# # Function to forecast number of fantasy points in a standard fantasy league

# def wr_forecast_f(player_name):
#     wr_model = ARIMA(player(player_name).FantPt, order = (2,1,0))
#     model_fit = wr_model.fit()
#     forecast = model_fit.forecast(steps = 1)

#     return print(forecast)
```

```
[18]: import warnings

warnings.filterwarnings('ignore')

forecast_f('Tom Brady')
```

```
9     295.490167
dtype: float64
```

```
[19]: # Function to return players 2020 fantasy points
def actual_f(player_name):
    df = test_df[test_df['Player'] == player_name]
    actual = df['FantPt'].iloc[0]
    return actual
```

```
[20]: # Function to evaluate the ARIMA model using mean absolute error as our error_
      ↪ statistic
def evaluate_arima_model(player_name, arima_order):

    model = ARIMA(player(player_name).FantPt, order = arima_order)
    model_fit = model.fit()
    forecast = model_fit.forecast(steps = 1)

    mean_absolute_error = np.mean(abs((actual_f(player_name) - forecast.
      ↪iloc[0])))

    return mean_absolute_error
```

```
[21]: order = (10,1,1)

evaluate_arima_model('DeAndre Hopkins',order)
```

```
[21]: 40.94534203196821
```

```
[22]: # Grid Search Function to find parameters for ARIMA that minimize the Mean
      ↪ Absolute Error
def evaluate_models(player_name, p_values, d_values, q_values):

    best_score, best_cfg = float("inf"), None

    for p in p_values:
        for d in d_values:
            for q in q_values:
                order = (p,d,q)
                try:
                    mae = evaluate_arima_model(player_name, order)
                    if mae < best_score:
                        best_score, best_cfg = mae, order
                    print('ARIMA%s MAE=%.3f' % (order,mae))
                except:
                    continue
    print('Best ARIMA%s MAE=%.3f' % (best_cfg, best_score))
```

```
[23]: # Function we use to build a pipeline
def evaluate_models1(player_name, p_values, d_values, q_values):
    #dataset = dataset[:1420]
    best_score, best_cfg = float("inf"), None

    for p in p_values:
        for d in d_values:
            for q in q_values:
                order = (p,d,q)
                try:
                    mae = evaluate_arima_model(player_name, order)
                    if mae < best_score:
                        best_score, best_cfg = mae, order
                        #print('ARIMA%s MAE=%.3f' % (order,mae))
                except:
                    continue
    print('Player:%s MAE=%.3f' % (player_name, best_score))
    return best_cfg
```

2 Player Forecast Function. enter player name and it will return the project fantasy points for the 2021 season. Optimizes ARIMA for each player as a QB's career trajectory is vastly different from a Running Back

```
[30]: def forecast_f_2021(player_name):
    p_values = [0, 1, 2, 3, 4, 5]
    d_values = range(0, 3)
    q_values = range(0, 10)

    best_cfg = evaluate_models1(player_name, p_values, d_values, q_values)
    rb_model = ARIMA(player(player_name).FantPt, order = best_cfg)
    model_fit = rb_model.fit()
    forecast = model_fit.forecast(steps = 2).iloc[1].round(decimals = 2)

    return print(player_name + ' Is projected: ' + str(forecast) + ' Fantasy_
    ↪points for the 2021 season. ')
```

```
[25]: # Forecast Function for the pipeline

def forecast_f_2021_actual(player_name):
    p_values = [0, 1, 2, 3, 4, 5]
    d_values = range(0, 3)
    q_values = range(0, 5)

    best_cfg = evaluate_models1(player_name, p_values, d_values, q_values)
    rb_model = ARIMA(player(player_name).FantPt, order = best_cfg)
    model_fit = rb_model.fit()
    forecast = model_fit.forecast(steps = 2).iloc[1].round(decimals = 2)

    return forecast
```

```
[26]: # Projected top 8 picks according to FantasyPros.com
projected_top_8 = ['Christian McCaffrey', 'Dalvin Cook', 'Derrick Henry', 'Saquon_
    ↪Barkley', 'Alvin Kamara', 'Ezekiel Elliott', 'Aaron Jones', 'Nick Chubb']

# Empty DF
forecast_df = pd.DataFrame()
# Empty Forecast List we will use in the loop below
forecast_list = []

#def rank_players(player_list):
for players in projected_top_8:
```

```

# Obtain Forecast and Append to list
player_forecast_2021 = forecast_f_2021_actual(players)
forecast_list.append(player_forecast_2021)

# Create name and Projected Fantasy Points DataFrame
forecast_df['Name'] = projected_top_8
forecast_df['Projected 2021 Fantasy Points'] = forecast_list

```

```

Player:Christian McCaffrey MAE=15.180
Player:Dalvin Cook MAE=55.717
Player:Derrick Henry MAE=98.393
Player:Saquon Barkley MAE=182.995
Player:Alvin Kamara MAE=9.394
Player:Ezekiel Elliott MAE=18.833
Player:Aaron Jones MAE=2.512
Player:Nick Chubb MAE=5.000

```

```

[27]: # View the DataFrame
forecast_df

```

```

[27]:
      Name  Projected 2021 Fantasy Points
0  Christian McCaffrey                88.18
1      Dalvin Cook                114.09
2      Derrick Henry                245.73
3      Saquon Barkley                295.00
4      Alvin Kamara                233.23
5      Ezekiel Elliott                234.64
6      Aaron Jones                 250.75
7      Nick Chubb                 197.00

```

```

[28]: # Sort DataFrame by Projected Fantasy Points
forecast_df = forecast_df.sort_values('Projected 2021 Fantasy Points',
↪ascending = False)

```

```

[29]: # Final Bar Plot showing project fantasy points
font = {'fontname': 'Georgia'}

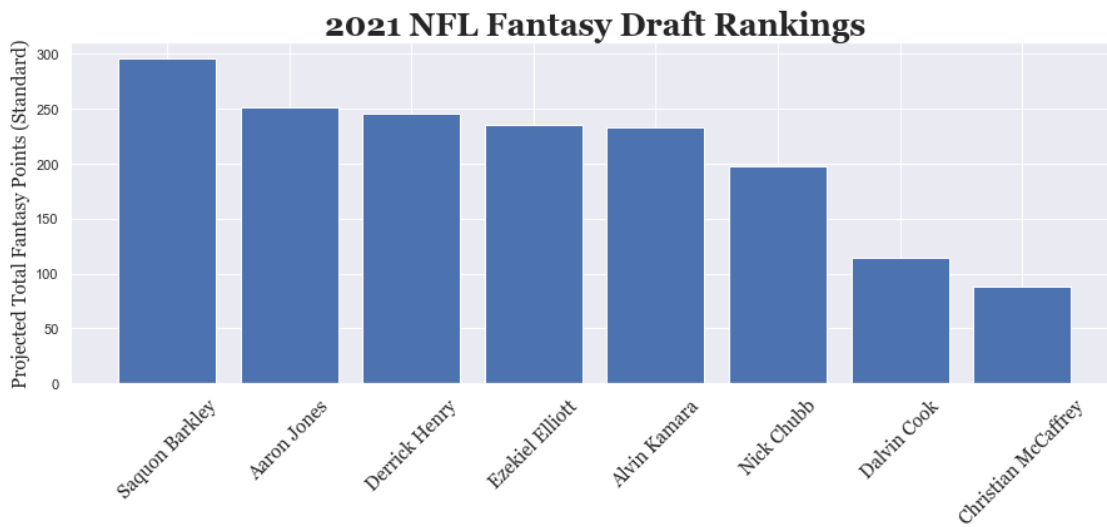
plt.figure(figsize = (15,5))

```



```
plt.bar(forecast_df['Name'], forecast_df['Projected 2021 Fantasy Points'] )
plt.xticks(rotation = 45, fontsize = 15, **font)
plt.title('2021 NFL Fantasy Draft Rankings',fontsize = 25, fontweight = 'bold',**font)
plt.ylabel('Projected Total Fantasy Points (Standard)',fontsize = 15,**font)
```

[29]: Text(0, 0.5, 'Projected Total Fantasy Points (Standard)')



[]: