```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import ttest ind
regular stats = pd.read csv('NBA Advanced Cleaned Stats.csv')
advanced stats = pd.read csv('nba stats cleaned.csv')
stats = pd.merge(regular stats, advanced stats, on='Player-
additional', suffixes=('_trad', '_adv'))
stats.head()
   Rk trad
                 Player trad Pos trad Age trad Tm trad G trad
MP trad
         1
           Precious Achiuwa
                                     C
                                              23
                                                     T<sub>0</sub>R
                                                               55
1140
                                                     MEM
         2
                Steven Adams
                                     C
                                              29
                                                               42
1
1133
         3
                                     C
                                              25
                                                     MIA
                                                               75
2
                 Bam Adebayo
2598
         4
                Ochai Agbaji
                                    SG
                                              22
                                                     UTA
                                                               59
3
1209
         5
                                    PF
                                              22
4
                Santi Aldama
                                                     MEM
                                                               77
1682
    PER
        TS%
                 3PAr ...
                              FT%
                                    0RB
                                         DRB
                                              TRB AST STL
                                                              BLK TOV
PF \
0 15.2 0.554 0.267 ... 0.702
                                    100
                                         228
                                              328
                                                    50
                                                         31
                                                               30
                                                                    59
102
                0.004 ...
                            0.364
                                              485
                                                                    79
1 17.5
         0.564
                                    214
                                         271
                                                    97
                                                         36
                                                               46
98
2 20.1
                0.011 ...
                            0.806
                                    184
                                         504
                                              688
                                                   240
         0.592
                                                         88
                                                               61
                                                                   187
208
3
    9.5
         0.561
                0.591 ...
                            0.812
                                     43
                                          78
                                              121
                                                    67
                                                         16
                                                               15
                                                                    41
99
         0.591 0.507 ...
4 13.9
                            0.750
                                     85
                                         286
                                              371
                                                    97
                                                         45
                                                               48
                                                                    60
143
    PTS
0
    508
1
    361
2
   1529
3
    467
    696
[5 rows x 61 columns]
sf = stats[stats['Pos trad']=='SF']
sf.reset index()
```

```
sf = sf.rename(columns={'Player':'Player Name'})
sf.head()
    Rk x
                  Player Name Pos x Unnamed: 3 Age x Tm x G x GS
MP_x
     \
13
      14
                                 SF
                                                   21
                                                                 3
                   OG Anunoby
                                           NaN
                                                       SAS
                                                              17
320
      16
                  Deni Avdija
                                 SF
                                           NaN
                                                   35
                                                       MIN
15
                                                              67
                                                                  66
2029
19
      20 Patrick Baldwin Jr.
                                 SF
                                           NaN
                                                   34
                                                        PH0
                                                              47
                                                                  47
1672
      28
27
               Scottie Barnes
                                 SF
                                           NaN
                                                   21
                                                       DET
                                                              47
                                                                 3
750
36
      37
             MarJon Beauchamp
                                 SF
                                           NaN
                                                   29
                                                       HOU
                                                              36
                                                                   0
235
                            WS/48
                                                 0BPM
     FG
              0WS
                   DWS
                         WS
                                    Unnamed: 24
                                                       DBPM
                                                              BPM
                                                                   VORP
/
13
                   2.9
                        4.7
                             0.094
                                                  -0.3
     59
              1.8
                                            NaN
                                                         0.7 0.4
                                                                    1.5
15
                   2.3
    261
              0.0
                        2.3
                             0.054
                                            NaN
                                                 -2.8
                                                        0.5 - 2.3 - 0.1
19
         ... -0.1
                                                 -0.7 -1.4 -2.1
    483
                   0.3
                        0.1
                             0.029
                                            NaN
                                                                    0.0
27
    102
         ... 2.3
                   2.7
                        5.0
                             0.090
                                            NaN
                                                  0.5 - 0.1 \ 0.4
                                                                    1.6
                                                  -4.1 -0.6 -4.8 -0.5
36
     31 ... -0.4 0.8
                        0.4
                             0.027
                                            NaN
    Player-additional y
13
              anunooq01
15
              avdiide01
              baldwpa01
19
27
              barnesc01
              beaucma01
36
[5 rows x 61 columns]
top =
['MIL','BOS','PHI','DEN','MEM','CLE','NYK','SAC','PHO','BRK','LAC','GS
W','LAL','ATL','MIA']
bottom =
['PHI','CHO','SAS','HOU','POR','ORL','WAS','IND','UTA','DAL','OKC','CH
I','TOR','NOP','MIN']
top sf = sf[sf['Tm x'].isin(top)]
     Rk_x
                        Player Pos_x Unnamed: 3 Age_x Tm_x G_x GS
MP x \
     20 Patrick Baldwin Jr. SF
19
                                            NaN
                                                    34
                                                         PH0
                                                               47
                                                                   47
1672
```

44 686	44		Вι	ıddy E	Boehei	m	SF		NaN	26	ATL	54	0
67	67		Ste	erling	g Brow	n	SF		NaN	23	LAL	78	65
1879 68	68		Tro	y Bro	own Jr		SF		NaN	29	PH0	42	0
687 118 1181	117		De	eMar [DeRoza	n	SF		NaN	21	DEN	76	6
	FG		OWS	DWS	WS	WS/4	8 U	nnamed	l: 24	OBPM	DBPM	BPM	
VORP 19 0.0	483		-0.1	0.3	0.1	0.02	9		NaN	-0.7	-1.4	-2.1	
44 0.2	73		-0.2	0.0	-0.2	-0.09	1		NaN	-7.5	-2.0	-9.5	-
67	247		0.0	0.1	0.0	0.04	0		NaN	-7.7	4.3	-3.3	
0.0 68	132		1.0	2.0	2.9	0.07	5		NaN	-1.8	0.8	-1.0	
0.5 118 2.6	143		5.4	3.2	8.5	0.15	3		NaN	1.7	0.3	2.0	
2.0	Dl ave	ar ad	lditic	nal v	,								
19 44 67 68 118	,		balo boeh brow brow	lwpa01 nebu01 nst02 ntr01 ozde01	l l 2 l								
[5 rd	ows x	61 c	olumn	ıs]									
botto	om_sf=	sf[s	sf['Tm	n_x'].	.isin(botto	m)]						
	Rk_x			Play	yer Po	s_x U	nnam	ed: 3	Age_	x Tm_x	G_x	GS	MP_x
13	14		0G	Anunc	oby	SF		NaN	2	1 SAS	17	3	320
59 15	16		Deni	. Avdi	ija	SF		NaN	3	5 MIN	67	66	2029
261 36 31	37	MarJ	Ion Be	eaucha	amp	SF		NaN	2	9 H0U	36	0	235
40	40		Sac	ldiq E	Веу	SF		NaN	2	2 DAL	15	0	108
22 48 77	48	Lea	ndro	Bolma	aro	SF		NaN	2	5 CH0	32	8	468
13 . 15 .	(OWS 1.8 0.0	DWS 2.9 2.3 0.8	WS 4.7 2.3 0.4	WS/48 0.094 0.054 0.027		amed	: 24 NaN NaN NaN	OBPM -0.3 -2.8 -4.1	DBPM 0.7 0.5 -0.6	BPM 0.4 -2.3 -4.8	VORI 1.! -0.!	5 1

```
40
                                                0.9
         2.6
              1.2 3.9
                         0.087
                                         NaN
                                                     -1.2
                                                           -0.3
                                                                   0.9
                                                     -2.3 -14.7
48
        -0.4
              0.0 - 0.3 - 0.231
                                         NaN -12.4
                                                                  -0.2
    Player-additional y
13
               anunooq01
15
               avdijde01
36
               beaucma01
40
                 bevsa01
48
               bolmale01
[5 rows x 61 columns]
top sf.describe()
                                      Gx
                                                   GS
                                                               MP x
             Rk_x
                        Age_x
FG \
count
        47.000000
                    47.000000
                                47.000000
                                           47.000000
                                                         47.000000
47.000000
                    26.063830
                                51.510638
                                           19.106383
                                                       1077.893617
mean
       299.042553
167.744681
       138.386503
std
                     4.560722
                                23.366524
                                           26.369642
                                                        755.049933
146.605722
                    19.000000
                                 4.000000
                                            0.000000
                                                         20.000000
min
        20.000000
3.000000
25%
       200.500000
                    22.500000
                                31.000000
                                            1.000000
                                                        402.000000
54.500000
50%
       311.000000
                    26,000000
                                61,000000
                                             4.000000
                                                       1042,000000
143.000000
75%
       399.000000
                    28.000000
                                70.500000
                                           25.500000
                                                       1745.000000
248.500000
                    36,000000
                                                       2382,000000
       526.000000
                                82.000000
                                           78.000000
max
587.000000
                FGA
                            FG%
                                         3P
                                                     3PA
                                                         . . .
                                                                Unnamed:
19 \
         47.000000
                     47.000000
                                  47.000000
                                               47.000000
count
0.0
        357.000000
                      0.451191
                                  58.489362
                                              159.085106
mean
NaN
std
        305.258835
                      0.074184
                                  60.520522
                                              149.747879
NaN
          9.000000
                      0.259000
                                   0.000000
min
                                                2.000000
NaN
25%
        142.500000
                      0.411500
                                   9.000000
                                               35.500000
NaN
50%
        288.000000
                      0.453000
                                  44.000000
                                              127.000000
NaN
                                             221.000000
75%
        469.500000
                      0.491000
                                  83.500000
NaN
       1252.000000
                      0.638000
                                 301.000000
                                              731.000000
max
```

```
NaN
             0WS
                         DWS
                                      WS
                                              WS/48
                                                      Unnamed: 24
OBPM
       47.000000
                   47.000000
                              47.000000
                                          47.000000
                                                               0.0
count
47.000000
mean
        0.953191
                    1.017021
                               1.957447
                                           0.076681
                                                               NaN
1.378723
std
                    1.015814
                                2.367419
                                           0.095830
                                                               NaN
        1.499689
4.224099
       -0.500000
                    0.000000
                               -0.200000
                                           -0.288000
                                                               NaN -
min
12,700000
25%
        0.050000
                    0.200000
                                0.200000
                                           0.041000
                                                               NaN
2.850000
                    0.700000
50%
        0.400000
                                1.000000
                                           0.068000
                                                               NaN
1.900000
75%
        1.050000
                    1.700000
                               2.900000
                                           0.112500
                                                               NaN
0.100000
        6.200000
                    4.300000
                               10.500000
                                           0.367000
                                                               NaN
max
17.000000
            DBPM
                         BPM
                                    VORP
       47.000000
                   47.000000
                              47.000000
count
       -0.168085
                   -1.542553
                                0.300000
mean
std
        2.265824
                    5.640218
                                1.028295
min
       -8.000000
                  -20.700000
                               -0.700000
25%
       -1.300000
                   -3.400000
                               -0.300000
50%
       -0.500000
                   -1.900000
                                0.000000
75%
        0.750000
                    0.050000
                                0.400000
        9.300000
                   17.800000
                                5.100000
max
[8 rows x 53 columns]
top sf = top sf[regular stats['MP']>1700]
/var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/
ipykernel 71183/387449806.py:1: UserWarning: Boolean Series key will
be reindexed to match DataFrame index.
  top sf = top sf[regular stats['MP']>1700]
                    Player Pos x Unnamed: 3 Age x Tm x G x
     Rk_x
                                                                 GS
                                                                     MP x
FG \
127
      126
             Tyler Dorsey
                              SF
                                         NaN
                                                  23
                                                      LAL
                                                            76
                                                                 45
                                                                     1860
200
              Josh Minott
337
      336
                              SF
                                         NaN
                                                  27
                                                      MIA
                                                            71
                                                                 49
                                                                     2077
254
378
      377 Eugene Omoruyi
                                         NaN
                                                  22
                                                      SAC
                                                            80
                                                                 78
                                                                     2382
                              SF
354
          0WS
               DWS
                      WS
                          WS/48 Unnamed: 24
                                               OBPM
                                                      DBPM
                                                              BPM
                                                                  VORP \
```

337		0.1 0.1 0.1	0.0 0.2 0.5		0.36 0.14 0.04	6		NaN NaN NaN	17.0 -1.9 -3.5	0.	4 -	17.8 -1.6 -4.4	0.0 0.0 -0.4
127 337 378	Playe	er-ado	dors mind	onal_y sety01 otjo01 oueu01									
[3 rov	ws x	61 co	olumr	ıs]									
bottor	m_sf.	desci	ribe(()									
		RI	<_x	Ag	ge_x		G_x		G:	S		MP_x	
FG \ count 52.000		2.000	900	52.000	0000	52.00	90000	52.	00000	9	52.0	900000	
mean 190.7	250).6346)	515	25.057	7692	50.0	19231	23.	96153	8 11	15.0	976923	
std		.9932	275	4.117	7935	20.62	22176	28.	34241	7 7	75.0	912482	
152.60 min 6.0000	14	, 1.0000	000	19.000	0000	6.00	90000	0.	00000	9	33.0	900000	
25%	126	5.7500	900	22.000	0000	33.50	90000	2.	00000	9 4	65.5	500000	
66.500 50% 148.00	232	2.5000	900	24.500	0000	54.00	90000	7.	00000	9	73.0	900000	
75% 284.2!	370	.7500	900	27.256	0000	65.50	90000	53.	25000	9 17	13.0	900000	
max 542.00	528	3.0000	900	37.000	0000	82.00	90000	80.	00000	9 26	78.0	900000	
			FGA		FG%		3	P		3PA		. Unn	amed:
19 \ count 0.0	5	52.000	9000	52.00	0000	52	. 00000	0	52.00	9000			
mean NaN	41	2.980	9769	0.47	72635	53	.88461	.5 1	155.65	3846			
std NaN	33	37.086	5914	0.05	57731	53	. 29078	7 1	.48.93	6739			
min NaN	1	2.000	0000	0.36	59000	2	.00000	0	5.000	0000			
25% NaN	12	28.750	0000	0.44	10500	7	. 75000	0	28.000	9000			
50% NaN	36	9.500	9000	0.45	57000	35	. 50000	0 1	13.00	9000			
75% NaN	61	10.75	0000	0.50	1500	80	. 00000	0 2	223.250	9000			
max NaN	114	12.000	0000	0.64	10000	212	. 00000	0 5	62.00	9000			

	0WS	DWS	WS	WS/48	B Unnamed:	24				
OBPM \ count 52.000	9000 52.0	00000 52	.000000	52.000000) 0	.0				
52.000000 mean 0.905	5769 1.1	42308 2	. 059615	0.064192	2 N	laN -				
1.401923 std 1.214	4038 0.9	31233 1.	. 923055	0.068500) N	laN				
3.030530 min -0.800	9000 0.0	00000 -0	.300000	-0.231000) N	laN -				
12.400000 25% 0.000	9000 0.3	00000 0	. 200000	0.046000) N	laN -				
3.000000 50% 0.700	9000 1.0	50000 1	. 950000	0.080006) N	laN -				
1.100000 75% 1.500	9000 1.8	25000 3	. 425000	0.095250) \	laN				
0.025000 max 4.900	9000 3.4	00000 7	. 600000	0.194000) N	laN				
DBPM BPM VORP count 52.000000 52.000000 52.000000 mean -0.278846 -1.684615 0.348077 std 1.218056 3.527894 0.821164 min -2.900000 -14.700000 -1.100000 25% -0.750000 -3.425000 -0.100000 50% -0.400000 -1.200000 0.150000 75% 0.500000 -0.300000 0.525000 max 3.500000 6.700000 3.500000 [8 rows x 53 columns] bottom_sf = bottom_sf[regular_stats['MP']>1700] /var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/ ipykernel_71183/2020337568.py:1: UserWarning: Boolean Series key will be reindexed to match DataFrame index.										
bottom_sf = Rk_x	<u> </u>	_			CTm_x G_x	GS	MP_x			
FG \ 15 16	Deni Avd	ija SF	N	laN 35	5 MIN 67	66	2029			
261 230 229 Bi 412	randon Ing	ram SF	N	laN 22	2 IND 56	56	1883			
456 455 467	Xavier Sn	eed SF	N	laN 23	B POR 62	62	2171			
0W9		S WS/48 3 0.054	Unnamed:	24 OBPM NaN -2.8			-			

```
230
         1.8 1.7 3.5 0.108
                                              2.2 -0.4 1.7
                                        NaN
                                                               1.4
         0.1 0.0 0.1 0.086
                                        NaN -3.1 -0.4 -3.6
456 ...
                                                               0.0
     Player-additional y
15
               avdijde01
230
               ingrabr01
456
               sneedxa01
[3 rows x 61 columns]
#significance tests
#PTS
stat, p value = ttest ind(top sf['PTS'], bottom sf['PTS'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in points scored.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in points scored.")
Test Statistic: -1.791258483153853
P-Value: 0.14964091532573953
Fail to reject the null hypothesis. There is no significant difference
in points scored.
#rebounds *************
stat, p value = ttest ind(top sf['TRB'], bottom sf['TRB'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in total rebounds.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in total rebounds.")
Test Statistic: 7.120551045460683
P-Value: 0.0031272607269830894
Reject the null hypothesis. There is a significant difference in total
rebounds.
```

```
#assists
stat, p value = ttest ind(top sf['AST'], bottom sf['AST'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in assists.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in assists.")
Test Statistic: -3.373296340413551
P-Value: 0.07668262409536224
Fail to reject the null hypothesis. There is no significant difference
in assists.
#turnovers *************
stat, p value = ttest ind(top sf['TOV'], bottom sf['TOV'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in turnovers.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in turnovers.")
Test Statistic: -4.407261060653622
P-Value: 0.013595756736915786
Reject the null hypothesis. There is a significant difference in
turnovers.
#FG%
stat, p value = ttest ind(top sf['FG%'], bottom sf['FG%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in field goal percentage.")
```

```
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in field goal percentage.")
Test Statistic: -0.28712845225021355
P-Value: 0.792165623355615
Fail to reject the null hypothesis. There is no significant difference
in field goal percentage.
#FT%
stat, p value = ttest ind(top sf['FT%'], bottom sf['FT%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in free throw %.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in free throw %.")
Test Statistic: -1.4628881276098042
P-Value: 0.23573290335082125
Fail to reject the null hypothesis. There is no significant difference
in free throw %.
#threes
stat, p value = ttest ind(top sf['3P%'], bottom sf['3P%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in threes.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in threes.")
Test Statistic: -0.27023200276876314
P-Value: 0.8063209767115538
Fail to reject the null hypothesis. There is no significant difference
in threes.
#steals
stat, p value = ttest ind(top sf['STL'], bottom sf['STL'],
```

```
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in steals.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in steals.")
Test Statistic: -0.3382120517773775
P-Value: 0.7639899588339132
Fail to reject the null hypothesis. There is no significant difference
in steals.
#blocks
stat, p value = ttest ind(top sf['BLK'], bottom sf['BLK'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in blocks.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in blocks.")
Test Statistic: 1.4320031115163143
P-Value: 0.25061548700423797
Fail to reject the null hypothesis. There is no significant difference
in blocks.
#personal fouls
stat, p value = ttest ind(top sf['PF'], bottom sf['PF'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in personal fouls.")
else:
```

```
print("Fail to reject the null hypothesis. There is no significant
difference in personal fouls.")
Test Statistic: 0.9581680931685118
P-Value: 0.40910865701561083
Fail to reject the null hypothesis. There is no significant difference
in personal fouls.
#offensive rebounds **************
stat, p value = ttest ind(top sf['ORB'], bottom sf['ORB'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in offensive rebounds.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in offensive rebounds.")
Test Statistic: 5.08812427331425
P-Value: 0.012903219189329136
Reject the null hypothesis. There is a significant difference in
offensive rebounds.
#efg%
stat, p value = ttest ind(top sf['eFG%'], bottom sf['eFG%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in eFG%.")
    print("Fail to reject the null hypothesis. There is no significant
difference in eFG%.")
Test Statistic: -0.066592715821201
P-Value: 0.9501052594481804
Fail to reject the null hypothesis. There is no significant difference
in eFG%.
#2P%
stat, p value = ttest ind(top sf['2P%'], bottom sf['2P%'],
equal var=False)
```

```
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in 2P%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in 2P%.")
Test Statistic: -0.1095282862900281
P-Value: 0.920085898362465
Fail to reject the null hypothesis. There is no significant difference
in 2P%.
#Defensive rebounds *******
stat, p_value = ttest_ind(top_sf['DRB'], bottom_sf['DRB'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in defensive rebounds.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in defensive rebounds.")
Test Statistic: 7.673618203378693
P-Value: 0.0017918062361505177
Reject the null hypothesis. There is a significant difference in
defensive rebounds.
#PER
stat, p_value = ttest_ind(top_sf['PER'], bottom sf['PER'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in PER.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in PER.")
```

```
Test Statistic: 1.0446087901944998
P-Value: 0.39271913906295447
Fail to reject the null hypothesis. There is no significant difference
in PER.
#true shooting
stat, p value = ttest ind(top sf['TS%'], bottom sf['TS%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in TS%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in TS%.")
Test Statistic: 0.4551615564047948
P-Value: 0.6807830391044389
Fail to reject the null hypothesis. There is no significant difference
in TS%.
#Usage Rate
stat, p value = ttest ind(top sf['USG%'], bottom sf['USG%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:
    print("Reject the null hypothesis. There is a significant
difference in USG%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in USG%.")
Test Statistic: 0.3253206099316201
P-Value: 0.7653328578957281
Fail to reject the null hypothesis. There is no significant difference
in USG%.
#Win Shares
stat, p value = ttest ind(top sf['WS'], bottom sf['WS'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
```

```
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in win shares.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in win shares.")
Test Statistic: -1.6234424267103007
P-Value: 0.24093502655161894
Fail to reject the null hypothesis. There is no significant difference
in win shares.
#Plus/Minus
stat, p_value = ttest_ind(top sf['BPM'], bottom sf['BPM'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in plus/minus.")
    print("Fail to reject the null hypothesis. There is no significant
difference in plus/minus.")
Test Statistic: 0.7448631363348374
P-Value: 0.5275451436095321
Fail to reject the null hypothesis. There is no significant difference
in plus/minus.
#VORP
stat, p value = ttest ind(top sf['VORP'], bottom sf['VORP'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in VORP.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in VORP.")
Test Statistic: -1.1283296297199523
P-Value: 0.36299505201722154
```

```
Fail to reject the null hypothesis. There is no significant difference
in VORP.
#3PAr
stat, p value = ttest ind(top sf['3PAr'], bottom sf['3PAr'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in 3PAr.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in 3PAr.")
Test Statistic: -0.5745640055064876
P-Value: 0.6010338010087608
Fail to reject the null hypothesis. There is no significant difference
in 3PAr.
#FTr
stat, p value = ttest ind(top sf['FTr'], bottom sf['FTr'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in FTr.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in FTr.")
Test Statistic: -1.4949728104292632
P-Value: 0.2500205834580632
Fail to reject the null hypothesis. There is no significant difference
in FTr.
#TRB%
stat, p value = ttest ind(top sf['TRB%'], bottom sf['TRB%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
```

```
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in TRB%.")
    print("Fail to reject the null hypothesis. There is no significant
difference in TRB%.")
Test Statistic: 1.2229371288986766
P-Value: 0.29286106573819953
Fail to reject the null hypothesis. There is no significant difference
in TRB%.
#0RB%
stat, p value = ttest ind(top sf['ORB%'], bottom sf['ORB%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:
    print("Reject the null hypothesis. There is a significant
difference in ORB%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in ORB%.")
Test Statistic: 2.452063837985608
P-Value: 0.09735057871545016
Fail to reject the null hypothesis. There is no significant difference
in ORB%.
#DRB%
stat, p value = ttest ind(top sf['DRB%'], bottom sf['DRB%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in DRB%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in DRB%.")
Test Statistic: -0.014754623763089749
P-Value: 0.9889390846688072
Fail to reject the null hypothesis. There is no significant difference
in DRB%.
```

```
#AST%
stat, p value = ttest ind(top sf['AST%'], bottom sf['AST%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in AST%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in AST%.")
Test Statistic: -2.6608821867764614
P-Value: 0.07871101030995326
Fail to reject the null hypothesis. There is no significant difference
in AST%.
#STL%
stat, p value = ttest ind(top sf['STL%'], bottom sf['STL%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in STL%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in STL%.")
Test Statistic: 0.6038772616713557
P-Value: 0.582554292759859
Fail to reject the null hypothesis. There is no significant difference
in STL%.
#BI K%
stat, p value = ttest ind(top sf['BLK%'], bottom sf['BLK%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in BLK%.")
```

```
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in BLK%.")
Test Statistic: 0.30499714066520933
P-Value: 0.7885210544983394
Fail to reject the null hypothesis. There is no significant difference
in BLK%.
#T0V%
stat, p value = ttest ind(top sf['TOV%'], bottom sf['TOV%'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p_value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in TOV%.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in TOV%.")
Test Statistic: -2.6425360276564973
P-Value: 0.07053587531728506
Fail to reject the null hypothesis. There is no significant difference
in TOV%.
#OWS
stat, p value = ttest ind(top sf['OWS'], bottom sf['OWS'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in OWS.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in OWS.")
Test Statistic: -0.9131682328565596
P-Value: 0.4575491971033516
Fail to reject the null hypothesis. There is no significant difference
in OWS.
/var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/
ipykernel 71183/444259772.py:2: RuntimeWarning: Precision loss
```

```
occurred in moment calculation due to catastrophic cancellation. This
occurs when the data are nearly identical. Results may be unreliable.
  stat, p value = ttest ind(top sf['OWS'], bottom sf['OWS'],
equal var=False)
#DWS
stat, p value = ttest ind(top sf['DWS'], bottom sf['DWS'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:
    print("Reject the null hypothesis. There is a significant
difference in DWS.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in DWS.")
Test Statistic: -1.5625952885742485
P-Value: 0.24857340343788484
Fail to reject the null hypothesis. There is no significant difference
in DWS.
#WS/48
stat, p value = ttest ind(top sf['WS/48'], bottom sf['WS/48'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:
    print("Reject the null hypothesis. There is a significant
difference in WS/48.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in WS/48.")
Test Statistic: 1.0539130319282333
P-Value: 0.3976268475084289
Fail to reject the null hypothesis. There is no significant difference
in WS/48.
#OBPM
stat, p value = ttest ind(top sf['OBPM'], bottom sf['OBPM'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
```

```
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in OBPM.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in OBPM.")
Test Statistic: 0.7496038319971363
P-Value: 0.5234199786381961
Fail to reject the null hypothesis. There is no significant difference
in OBPM.
#DBPM
stat, p_value = ttest_ind(top_sf['DBPM'], bottom sf['DBPM'],
equal var=False)
alpha = 0.05
print(f'Test Statistic: {stat}')
print(f'P-Value: {p value}')
if p value < alpha:</pre>
    print("Reject the null hypothesis. There is a significant
difference in DBPM.")
else:
    print("Fail to reject the null hypothesis. There is no significant
difference in DBPM.")
Test Statistic: 0.33646329245522666
P-Value: 0.7572438453181312
Fail to reject the null hypothesis. There is no significant difference
in DBPM.
#Stats of significance: TOV, TRB, ORB, DRB
weights = \{'TOV': 0.25, 'TRB': 0.25, 'ORB': 0.25, 'DRB': 0.25\}
for stat in weights.keys():
    max value = sfs[stat].max()
    sfs.loc[:, stat + ' norm'] = sfs[stat]/max value
sfs.loc[:, 'Weighted Score'] = sum([sfs[stat + ' norm'] * weight for
stat, weight in weights.items()])
ranked sf = sfs.sort values('Weighted Score', ascending=False)
print(ranked sf[['Player','Weighted Score']].head(25))
                    Player Weighted Score
408
                                  0.776684
               Cam Reddish
96
               Amir Coffey
                                  0.746137
```

```
221
             Caleb Houstan
                                   0.742874
392
           Otto Porter Jr.
                                   0.736813
67
            Sterling Brown
                                   0.730085
461
                 Max Strus
                                   0.695419
190
          Tim Hardawav Jr.
                                   0.678223
310
              Caleb Martin
                                   0.665365
495
             Jabari Walker
                                   0.532698
180
             Javonte Green
                                   0.492434
503
             Yuta Watanabe
                                   0.489638
149
         Simone Fontecchio
                                   0.486385
378
            Eugene Omoruyi
                                   0.450185
337
               Josh Minott
                                   0.448499
478
     Juan Toscano-Anderson
                                   0.421273
128
             Luquentz Dort
                                   0.406768
511
                Jack White
                                   0.401645
336
             Justin Minaya
                                   0.389453
230
            Brandon Ingram
                                   0.370089
            Hamidou Diallo
121
                                   0.366111
127
              Tyler Dorsey
                                   0.364787
380
           Kelly Oubre Jr.
                                   0.347918
229
                Joe Ingles
                                   0.313076
456
              Xavier Sneed
                                   0.306511
15
               Deni Avdija
                                   0.303863
```

/var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/
ipykernel_71183/4022255224.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation:

https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

sfs.loc[:, stat + '_norm'] = sfs[stat]/max_value

/var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/ipykernel_71183/40222 55224.py:4: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation:

https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#
returning-a-view-versus-a-copy

sfs.loc[:, stat + ' norm'] = sfs[stat]/max value

/var/folders/wp/jqq6rhr97930vxndhx84wd8w0000gn/T/ipykernel_71183/40222 55224.py:4: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation:

https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
sfs.loc[:, stat + ' norm'] = sfs[stat]/max value
/var/folders/wp/jgg6rhr97930vxndhx84wd8w0000gn/T/ipykernel 71183/40222
55224.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#
returning-a-view-versus-a-copy
  sfs.loc[:, stat + ' norm'] = sfs[stat]/max value
/var/folders/wp/jgg6rhr97930vxndhx84wd8w0000gn/T/ipykernel 71183/40222
55224.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#
returning-a-view-versus-a-copy
  sfs.loc[:, 'Weighted Score'] = sum([sfs[stat + ' norm'] * weight for
stat, weight in weights.items()])
salaries = pd.read csv('Nba Player Salaries.csv')
salaries = salaries.rename(columns={'2022/2023': 'Salary'})
salaries['Salary'] = salaries['Salary'].replace('[\$,]', '',
regex=True).astype(int)
salaries.head()
   Player Id
                    Player Name
                                   Salary
                                              2023/2024
                                                            2024/2025
\
0
                  Stephen Curry 48070014
                                           $51,915,615
                                                         $55,761,217
           1
           2
                      John Wall 47345760
                                                    $0
                                                                  $0
1
2
             Russell Westbrook 47080179
                                                    $0
                                                                  $0
                   LeBron James 44474988
                                           $46,698,737 $50,434,636
                   Kevin Durant 44119845
                                           $47,649,433
                                                         $51,179,020
           5
    2024/2025.1
   $59,606,817
1
            $0
2
            $0
3
            $0
   $54,708,608
final sfs = pd.merge(ranked sf, salaries, on='Player Name')
final_sfs = final_sfs[['Player Name','Weighted_Score','Salary']]
final sfs = pd.merge(final sfs, sf[['Player
Name', 'Pos x', 'Tm x']], on='Player Name')
```

```
final sfs =
final sfs.rename(columns={'Pos x':'Position','Tm x':'Team'})
order = ['Player Name', 'Team', 'Position', 'Weighted Score',
'Salarv'l
final sfs = final sfs[order] if all(col in final sfs.columns for col
in order) else final sfs
final sfs
                                            Weighted Score
               Player Name Team Position
                                                                Salary
                                                   0.776684
0
               Cam Reddish
                             MIL
                                                               5954454
                                        SF
1
                                        SF
               Amir Coffey
                             T0R
                                                   0.746137
                                                               3395062
2
             Caleb Houstan
                             GSW
                                        SF
                                                   0.742874
                                                               2000000
3
            Sterling Brown
                                        SF
                             LAL
                                                   0.730085
                                                               3122602
4
                 Max Strus
                                        SF
                             HOU
                                                   0.695419
                                                               1815677
5
              Caleb Martin
                             WAS
                                        SF
                                                   0.665365
                                                               6479000
6
             Jabari Walker
                                        SF
                                                   0.532698
                             0RL
                                                               1017781
7
             Javonte Green
                             0KC
                                        SF
                                                   0.492434
                                                               1815677
8
                                        SF
             Yuta Watanabe
                            CH0
                                                   0.489638
                                                               1968175
9
        Simone Fontecchio
                             UTA
                                        SF
                                                   0.486385
                                                               3205128
10
           Eugene Omoruyi
                             SAC
                                        SF
                                                   0.450185
                                                               1013119
11
                                        SF
                                                   0.448499
               Josh Minott
                             MIA
                                                               1017781
12
    Juan Toscano-Anderson
                             GSW
                                        SF
                                                   0.421273
                                                               2133278
13
                             NYK
                                        SF
                                                   0.406768
                                                              15277778
             Luguentz Dort
14
                Jack White
                             B<sub>0</sub>S
                                        SF
                                                   0.401645
                                                                508891
15
                             NOP
                                        SF
             Justin Minaya
                                                   0.389453
                                                                 35096
16
            Brandon Ingram
                             IND
                                        SF
                                                   0.370089
                                                              31650600
            Hamidou Diallo
                                        SF
17
                             MEM
                                                   0.366111
                                                               5200000
18
              Tyler Dorsey
                             LAL
                                        SF
                                                   0.364787
                                                                201802
19
                Joe Ingles
                             NYK
                                        SF
                                                   0.313076
                                                               6479000
20
              Xavier Sneed
                             P<sub>0</sub>R
                                        SF
                                                   0.306511
                                                                102910
21
               Deni Avdija
                            MIN
                                        SF
                                                   0.303863
                                                               4916160
```

final_sfs.to_csv('sfs_final_list.csv',index=False)