

IEOR 142 Final Project Linear Regression

May 12, 2021

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

[2]: # compute out-of-sample R-squared using the test set
def OSR2(model, df_train, df_test, dependent_var):
    y_test = df_test[dependent_var]
    y_pred = model.predict(df_test)
    SSE = np.sum((y_test - y_pred)**2)
    SST = np.sum((y_test - np.mean(df_train[dependent_var]))**2)
    return 1 - SSE/SST
```

1 Model 2: Linear Regression

```
[3]: dataset = pd.read_csv('update1422.csv')
```

Most Valuable Player (MVP)

```
[160]: #mvp models
dataset = pd.read_csv('update142.csv')
train = dataset[dataset['Season'] <= 2017]
test = dataset[dataset['Season'] > 2017]

y_train = train['MVP']
y_test = test['MVP']
x_train = train.iloc[:,6:51]
x_test = test.iloc[:,6:51]
```

```
[5]: x_train.head(5)
```

```
[5]:
```

	G	GS	MP	FGM	FGA	FGP	X3PM	X3PA	X3PP	X2PM	...	TOVP	USG	\
0	22	0	7.4	0.8	1.9	0.405	0.2	0.5	0.500	0.5	...	16.4	17.6	
1	56	2	15.4	2.6	6.3	0.410	1.1	3.1	0.345	1.5	...	13.2	22.7	
2	50	0	15.9	2.3	6.4	0.356	0.8	3.0	0.275	1.5	...	13.0	22.7	
3	44	1	12.9	1.3	4.0	0.339	0.6	2.0	0.295	0.8	...	14.5	17.7	
4	57	22	22.4	2.8	7.2	0.390	1.2	3.5	0.350	1.6	...	12.7	18.0	

	OWS	DWS	WS	WS48	OBPM	DBPM	BPM	VORP
0	-0.2	0.2	0.0	-0.001	-6.6	0.0	-6.6	-0.2
1	0.4	0.8	1.2	0.065	-0.1	-0.4	-0.5	0.3
2	-0.4	0.7	0.3	0.020	-1.9	-1.1	-2.9	-0.2
3	0.2	0.5	0.7	0.063	-0.6	-0.5	-1.0	0.1
4	1.0	1.2	2.2	0.084	-0.2	-0.4	-0.6	0.5

[5 rows x 45 columns]

```
[6]: y_train.head(5)
```

```
[6]: 0    0.0
      1    0.0
      2    0.0
      3    0.0
      4    0.0
      Name: MVP, dtype: float64
```

```
[7]: x_train.columns
```

```
[7]: Index(['G', 'GS', 'MP', 'FGM', 'FGA', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM',
          'X2PA', 'X2PP', 'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST',
          'STL', 'BLK', 'TOV', 'PF', 'PTS', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP',
          'DRBP', 'TRBP', 'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
          'WS', 'WS48', 'OBPM', 'DBPM', 'BPM', 'VORP'],
          dtype='object')
```

```
[141]: #taking out the 4 additional awards plus categorical stats
        dependant_var = train.columns.difference(['Tm', 'Pos', 'Player', 'MVP',
        ↪ 'DPOY', 'ROY', 'SMOY', 'MIP'])
        dependant_var
```

```
[141]: Index(['AST', 'ASTP', 'Age', 'BLK', 'BLKP', 'BPM', 'DBPM', 'DRB', 'DRBP',
          'DWS', 'EFG', 'FGA', 'FGM', 'FGP', 'FTA', 'FTM', 'FTP', 'FTR', 'G',
          'GS', 'MP', 'OBPM', 'ORB', 'ORBP', 'OWS', 'PER', 'PF', 'PTS', 'STL',
          'STLP', 'Season', 'TOV', 'TOVP', 'TRB', 'TRBP', 'TS', 'USG',
          'Unnamed: 0', 'VORP', 'WS', 'WS48', 'X2PA', 'X2PM', 'X2PP', 'X3PA',
          'X3PAR', 'X3PM', 'X3PP'],
          dtype='object')
```

```
[142]: #creating list for regression
        dependant_var = "+".join(dependant_var)
```

```
[143]: dependant_var
```

```
[143]: 'AST+ASTP+Age+BLK+BLKP+BPM+DBPM+DRB+DRBP+DWS+EFG+FGA+FGM+FGP+FTA+FTM+FTP+FTR+G+G
        S+MP+OBPM+ORB+ORBP+OWS+PER+PF+PTS+STL+STLP+Season+TOV+TOVP+TRB+TRBP+TS+USG+Unnam
```

ed: 0+VORP+WS+WS48+X2PA+X2PM+X2PP+X3PA+X3PAR+X3PM+X3PP'

```
[144]: train.head(5)
```

```
[144]: Unnamed: 0      Player Season Pos Age  Tm  G  GS  MP  FGM  ...  \
0          1  A.J. Hammons  2017  C   24  DAL  22  0   7.4  0.8  ...
1          2    A.J. Price  2010  PG  23  IND  56  2  15.4  2.6  ...
2          3    A.J. Price  2011  PG  24  IND  50  0  15.9  2.3  ...
3          4    A.J. Price  2012  PG  25  IND  44  1  12.9  1.3  ...
4          5    A.J. Price  2013  PG  26  WAS  57  22  22.4  2.8  ...

      WS48  OBPM  DBPM  BPM  VORP  DPOY  ROY  SMOY  MIP  MVP
0 -0.001 -6.6   0.0 -6.6  -0.2   0.0  0.0   0.0  0.0  0.0
1  0.065 -0.1  -0.4 -0.5   0.3   0.0  0.0   0.0  0.0  0.0
2  0.020 -1.9  -1.1 -2.9  -0.2   0.0  0.0   0.0  0.0  0.0
3  0.063 -0.6  -0.5 -1.0   0.1   0.0  0.0   0.0  0.0  0.0
4  0.084 -0.2  -0.4 -0.6   0.5   0.0  0.0   0.0  0.0  0.0
```

[5 rows x 56 columns]

```
[145]: cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
               'FGM', 'FGA', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
               'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK',
               'TOV', 'PF', 'PTS', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
               'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS', 'WS48',
               'OBPM', 'DBPM', 'BPM', 'VORP']
```

```
[147]: #calculate VIF
import statsmodels.api as sm

from statsmodels.stats.outliers_influence import variance_inflation_factor

# The dataframe passed to VIF must include the intercept term
def VIF(df, columns):
    values = sm.add_constant(df[columns]).values
    num_columns = len(columns)+1
    vif = [variance_inflation_factor(values, i) for i in range(num_columns)]
    return pd.Series(vif[1:], index=columns)
```

```
[148]: VIF(train,cols)
```

```
[148]: Season      1.194476
Age          1.171180
G            2.391408
GS           1.641163
MP           19.132866
FGM          1569.891795
```

FGA	3525.506741
FGP	36.893949
X3PM	269.847081
X3PA	1029.538971
X3PP	2.360528
X2PM	639.446195
X2PA	2108.013946
X2PP	7.219283
EFG	26.918430
FTM	141.599957
FTA	60.443550
FTP	2.412699
ORB	148.398531
DRB	452.846363
TRB	949.698229
AST	9.618181
STL	5.474090
BLK	6.334876
TOV	10.020104
PF	4.023531
PTS	2106.101220
PER	82.580696
TS	11.568447
X3PAR	8.638536
FTR	3.819038
ORBP	352.290197
DRBP	650.737814
TRBP	1610.725851
ASTP	8.143810
STLP	5.133920
BLKP	7.322589
TOVP	4.740989
USG	11.597479
OWS	685.531693
DWS	254.394251
WS	1379.407787
WS48	32.905514
OBPM	2664.256508
DBPM	729.156559
BPM	4008.612810
VORP	12.315918

dtype: float64

```
[149]: #Due to the high VIF above we are going to remove FGA
depedent_var = train.columns.difference(['Tm', 'Pos', 'Player', 'MVP', '
↳ 'DPOY', 'ROY', 'SMOY', 'MIP', 'FGA'])
depedent_var
```

```
depedent_var = "+".join(depedent_var)
```

```
[150]: cols = [ 'Season', 'Age', 'G', 'GS', 'MP',  
               'FGM', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',  
               'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK',  
               'TOV', 'PF', 'PTS', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',  
               'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS', 'WS48',  
               'OBPM', 'DBPM', 'BPM', 'VORP']
```

```
[151]: VIF(train,cols)
```

```
[151]: Season      1.192306  
Age          1.170550  
G            2.390958  
GS           1.641136  
MP           19.116749  
FGM          1569.539979  
FGP           36.878469  
X3PM          269.318653  
X3PA           83.753328  
X3PP           2.356868  
X2PM          639.400745  
X2PA          112.193462  
X2PP           7.211029  
EFG           26.865980  
FTM           141.549555  
FTA           60.438245  
FTP            2.412662  
ORB           148.089320  
DRB           451.506745  
TRB           947.420101  
AST            9.616408  
STL            5.473669  
BLK            6.329678  
TOV           10.014365  
PF             4.022295  
PTS           2105.344255  
PER            82.536375  
TS            11.539527  
X3PAR          8.636201  
FTR            3.818829  
ORBP           352.189012  
DRBP           650.453957  
TRBP           1609.899949  
ASTP            8.136427  
STLP            5.124074  
BLKP            7.321098
```

```

TOVP          4.731929
USG           11.596995
OWS           683.923741
DWS           253.582976
WS            1375.251650
WS48          32.884934
OBPM          2663.571876
DBPM          729.062872
BPM           4008.019099
VORP          12.314954
dtype: float64

```

```

[152]: #removing pts due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGM', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS', 'WS48',
         'OBPM', 'DBPM', 'BPM', 'VORP']

```

```

[21]: VIF(train,cols)

```

```

[21]: Season          1.192514
Age                1.157173
G                  2.157620
GS                 3.608952
MP                 23.900449
FGM               2283.109243
FGP               31.824353
X3PM              291.274036
X3PA              89.004363
X3PP              2.237480
X2PM              1920.781309
X2PA              176.130874
X2PP              7.776078
EFG               19.930258
FTM               72.487559
FTA               75.538858
FTP               2.264529
ORB               276.497219
DRB               1197.834521
TRB               2330.703775
AST               11.951138
STL               6.287637
BLK               5.729287
TOV               14.439666
PF                3.709223

```

```

PER      101.732875
TS        8.920161
X3PAR     8.271101
FTR       3.310331
ORBP     327.809995
DRBP     705.569352
TRBP    1673.008052
ASTP      9.241032
STLP      5.030717
BLKP      6.481220
TOVP      3.568362
USG       13.940051
OWS     1779.961837
DWS       540.702192
WS      3496.352067
WS48      35.859572
OBPM     3167.753461
DBPM      709.863068
BPM     4514.127858
VORP      24.534486
dtype: float64

```

```

[22]: #removing bpm due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
          'FGM', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
          'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK',
          'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
          'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS', 'WS48',
          'OBPM', 'DBPM', 'VORP']

```

```

[23]: VIF(train,cols)

```

```

[23]: Season      1.191270
Age        1.156164
G          2.157619
GS         3.608781
MP        23.899740
FGM      2282.132921
FGP       31.819902
X3PM     291.245169
X3PA      88.987430
X3PP      2.237461
X2PM    1920.585170
X2PA     176.015777
X2PP       7.771931
EFG       19.928501
FTM       72.454841

```

```

FTA          75.471022
FTP          2.263889
ORB          276.380524
DRB          1197.362035
TRB          2329.768191
AST          11.951136
STL          6.284697
BLK          5.729277
TOV          14.439240
PF           3.709073
PER          101.630358
TS           8.911092
X3PAR        8.271061
FTR          3.304577
ORBP         327.529379
DRBP         705.241986
TRBP         1672.228619
ASTP         9.235713
STLP         5.030714
BLKP         6.474528
TOVP         3.562774
USG          13.922676
OWS          1779.322436
DWS          540.486595
WS           3494.802942
WS48         35.852381
OBPM         31.997043
DBPM         9.361493
VORP         24.522676
dtype: float64

```

```

[24]: #removing ws due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGM', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
         'OBPM', 'DBPM', 'VORP']

```

```

[25]: VIF(train,cols)

```

```

[25]: Season      1.191115
      Age         1.156049
      G           2.153862
      GS          3.607895
      MP          23.898944
      FGM         2281.864283

```


FGP	31.817768
X3PM	291.242649
X3PA	88.975874
X3PP	2.237320
X2PM	1920.166660
X2PA	175.970957
X2PP	7.771130
EFG	19.927337
FTM	72.454820
FTA	75.466325
FTP	2.263751
ORB	276.367278
DRB	1197.224624
TRB	2329.498754
AST	11.949438
STL	6.282586
BLK	5.729181
TOV	14.437050
PF	3.708898
PER	101.553367
TS	8.909877
X3PAR	8.269175
FTR	3.301326
ORBP	327.444951
DRBP	704.904861
TRBP	1671.389859
ASTP	9.234727
STLP	5.030692
BLKP	6.473371
TOVP	3.562668
USG	13.912545
OWS	20.766398
DWS	6.221203
WS48	35.848544
OBPM	31.942630
DBPM	9.354186
VORP	24.440627

dtype: float64

```
[26]: #removing trb due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
        'FGM', 'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
        'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
        'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
        'OBPM', 'DBPM', 'VORP']
VIF(train,cols)
```

```
[26]: Season      1.191098
      Age        1.156027
      G          2.151490
      GS         3.605488
      MP         23.887734
      FGM        2281.758838
      FGP         31.777961
      X3PM        291.121029
      X3PA        88.926900
      X3PP         2.237278
      X2PM       1920.092467
      X2PA       175.970938
      X2PP         7.767782
      EFG        19.921656
      FTM        72.437155
      FTA        75.464105
      FTP         2.263644
      ORB        11.028235
      DRB        13.013680
      AST        11.939587
      STL         6.282178
      BLK         5.719172
      TOV        14.420354
      PF          3.707280
      PER       101.553365
      TS          8.909282
      X3PAR        8.250469
      FTR         3.299641
      ORBP       327.009506
      DRBP       704.484994
      TRBP      1670.350156
      ASTP         9.228112
      STLP         5.026954
      BLKP         6.460002
      TOVP         3.553561
      USG        13.911940
      OWS        20.766174
      DWS         6.221037
      WS48        35.845835
      OBPM        31.942169
      DBPM         9.342066
      VORP       24.440467
      dtype: float64
```

```
[27]: #removing fgm due to high VIF
      cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
               'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
```

```
'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP', 'TRBP',
'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
'OBPM', 'DBPM', 'VORP']
```

```
VIF(train,cols)
```

```
[27]: Season      1.191091
      Age         1.155668
      G           2.151457
      GS          3.605458
      MP          23.874730
      FGP         31.766296
      X3PM        81.536482
      X3PA        86.184756
      X3PP        2.234934
      X2PM        171.871076
      X2PA        173.805932
      X2PP        7.753346
      EFG         19.831132
      FTM         72.424124
      FTA         75.463811
      FTP         2.263583
      ORB         11.027267
      DRB         13.007214
      AST         11.939387
      STL         6.278886
      BLK         5.717805
      TOV         14.420343
      PF          3.706706
      PER         101.518087
      TS          8.893118
      X3PAR        8.250464
      FTR          3.299591
      ORBP        326.981006
      DRBP        704.430155
      TRBP        1670.288967
      ASTP        9.207796
      STLP        5.024333
      BLKP        6.458940
      TOVP        3.547873
      USG         13.911815
      OWS         20.750167
      DWS         6.216409
      WS48        35.843736
      OBPM        31.936323
      DBPM        9.341701
      VORP        24.439215
```

dtype: float64

```
[28]: #removing trbp due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PA', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
         'OBPM', 'DBPM', 'VORP']
VIF(train,cols)
```

```
[28]: Season      1.190743
      Age        1.154935
      G          2.150657
      GS         3.604758
      MP        23.874285
      FGP        31.761892
      X3PM       81.251721
      X3PA       85.676643
      X3PP        2.232876
      X2PM       169.497259
      X2PA       171.227676
      X2PP         7.752946
      EFG        19.831128
      FTM        72.412926
      FTA        75.434079
      FTP         2.263343
      ORB        10.843960
      DRB        12.914462
      AST        11.938813
      STL         6.276723
      BLK         5.688668
      TOV        14.361535
      PF          3.655416
      PER        96.311802
      TS          8.890759
      X3PAR       8.241833
      FTR         3.295449
      ORBP       11.112562
      DRBP        7.624789
      ASTP        9.206936
      STLP        5.023912
      BLKP        6.458082
      TOVP        3.540283
      USG        12.927476
      OWS        20.625135
      DWS         6.190195
```

```

WS48      33.898806
OBPM      31.725791
DBPM      9.254250
VORP      24.296651
dtype: float64

```

```

[29]: #removing X2PA due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'PER', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
         'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[29]: Season      1.190738
      Age         1.154452
      G           2.140194
      GS          3.602211
      MP          22.343352
      FGP         31.691341
      X3PM        79.032802
      X3PA        83.297459
      X3PP         2.232309
      X2PM        15.664618
      X2PP         6.991430
      EFG         19.054164
      FTM         67.416949
      FTA         73.037851
      FTP          2.253342
      ORB         10.836093
      DRB         12.670390
      AST         11.906861
      STL          6.250501
      BLK          5.625577
      TOV         14.345090
      PF           3.629666
      PER         94.908807
      TS           8.780881
      X3PAR        8.163553
      FTR          3.286643
      ORBP        11.101929
      DRBP         7.534487
      ASTP         9.162106
      STLP         4.961063
      BLKP         6.379067
      TOVP         3.478914

```

```

USG      12.680292
OWS      18.516362
DWS      6.183209
WS48     33.747353
OBPM     31.184750
DBPM     9.171278
VORP     24.201876
dtype: float64

```

```

[30]: #removing PER due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGP', 'X3PM', 'X3PA', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
         'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[30]: Season      1.182011
Age          1.150861
G            2.139219
GS           3.602163
MP           22.275823
FGP          31.019189
X3PM         79.032557
X3PA         83.237394
X3PP         2.225294
X2PM         15.462851
X2PP         6.926645
EFG          18.989398
FTM          66.608099
FTA          72.370881
FTP          2.252218
ORB          10.834954
DRB          12.608040
AST          11.905542
STL          6.229601
BLK          5.495950
TOV          14.343970
PF           3.520283
TS           7.707021
X3PAR        6.758435
FTR          3.261893
ORBP         10.888835
DRBP         5.823802
ASTP         8.224600
STLP         3.552044

```

```

BLKP      5.341370
TOVP      3.459957
USG       5.939072
OWS       17.674266
DWS       5.997991
WS48      23.222693
OBPM      15.253435
DBPM      9.116896
VORP      22.931276
dtype: float64

```

```

[31]: #removing X3PA due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'FGP', 'X3PM', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTM', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
         'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[31]: Season      1.176277
Age          1.150852
G            2.136088
GS           3.580865
MP           21.741809
FGP          29.424692
X3PM         8.565349
X3PP         2.214324
X2PM         15.450172
X2PP         5.050942
EFG          18.543103
FTM          66.607056
FTA          72.192295
FTP          2.252135
ORB          10.735525
DRB          12.579199
AST          11.905477
STL          6.218975
BLK          5.493915
TOV          14.343678
PF           3.518936
TS           7.683229
X3PAR        6.627916
FTR          3.261631
ORBP         10.785946
DRBP         5.817433
ASTP         8.201403

```

```

STLP      3.551842
BLKP      5.325940
TOVP      3.437009
USG       5.933374
OWS       17.404690
DWS       5.992156
WS48      23.210584
OBPM      15.247378
DBPM      9.103365
VORP      22.917772
dtype: float64

```

```

[32]: #removing FTM due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
        'FGP', 'X3PM', 'X3PP', 'X2PM', 'X2PP',
        'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
        'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
        'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[32]: Season      1.176277
      Age         1.147968
      G           2.136082
      GS          3.576532
      MP          21.741141
      FGP         29.118624
      X3PM        8.352450
      X3PP        2.212089
      X2PM        15.440217
      X2PP        5.050904
      EFG         17.900881
      FTA         9.351436
      FTP         1.634758
      ORB         10.549852
      DRB         12.568472
      AST         11.902180
      STL         6.199054
      BLK         5.464350
      TOV         14.343330
      PF          3.517155
      TS          7.675291
      X3PAR       6.624863
      FTR         3.142224
      ORBP       10.784063
      DRBP        5.814025
      ASTP        8.182898

```



```

STLP      3.551730
BLKP      5.311123
TOVP      3.413876
USG       5.930774
OWS       17.225947
DWS       5.877662
WS48      23.114179
OBPM      15.241474
DBPM      9.103190
VORP      22.887696
dtype: float64

```

```

[33]: #removing FGP
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
        'X3PM', 'X3PP', 'X2PM', 'X2PP',
        'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
        'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS', 'WS48',
        'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[33]: Season      1.176191
      Age         1.147318
      G           2.133058
      GS          3.574171
      MP          21.676414
      X3PM        7.363673
      X3PP        2.038686
      X2PM        14.971425
      X2PP        4.036321
      EFG         5.266394
      FTA         9.349609
      FTP         1.625006
      ORB         10.282187
      DRB         12.553260
      AST         11.845687
      STL         6.198519
      BLK         5.460107
      TOV         14.335889
      PF          3.498074
      TS          7.675265
      X3PAR       4.688249
      FTR         3.139627
      ORBP        10.589668
      DRBP        5.795417
      ASTP        8.148626
      STLP        3.551492

```

```

BLKP      5.311048
TOVP      3.395314
USG       5.930508
OWS       17.103823
DWS       5.805003
WS48      23.000550
OBPM      15.227228
DBPM      9.088319
VORP      22.738849
dtype: float64

```

```

[34]: #removing WS48 due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
         'X3PM', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'OBPM', 'DBPM', 'VORP']
VIF(train,cols)

```

```

[34]: Season      1.176167
      Age        1.145824
      G          2.072139
      GS         3.573711
      MP        19.915127
      X3PM       7.148818
      X3PP       2.034893
      X2PM      14.439520
      X2PP       4.035960
      EFG       5.261874
      FTA       9.218844
      FTP       1.612970
      ORB      10.246025
      DRB      12.552687
      AST      11.574889
      STL       6.189365
      BLK       5.448306
      TOV      14.089805
      PF        3.479433
      TS        6.158846
      X3PAR     4.677784
      FTR       3.107208
      ORBP     9.017877
      DRBP     5.787690
      ASTP     8.069173
      STLP     3.394789
      BLKP     5.044777

```

```

TOVP      3.072474
USG       5.740124
OWS      14.866244
DWS       5.190462
OBPM      9.682280
DBPM      5.325734
VORP     20.186208
dtype: float64

```

```

[35]: #removing VORP due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS', 'MP',
        'X3PM', 'X3PP', 'X2PM', 'X2PP',
        'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
        'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        'OBPM', 'DBPM', ]
VIF(train,cols)

```

```

[35]: Season      1.164221
Age          1.144936
G            1.921076
GS           3.571630
MP          18.211867
X3PM         6.817424
X3PP         2.034562
X2PM        14.393876
X2PP         4.034296
EFG          5.257562
FTA          9.206455
FTP          1.601102
ORB         10.202223
DRB         12.086051
AST         11.419129
STL          5.758635
BLK          5.235173
TOV         14.082940
PF           3.261486
TS           5.873030
X3PAR        4.653031
FTR          3.106413
ORBP         9.017397
DRBP         5.772044
ASTP         8.064346
STLP         3.383403
BLKP         5.030034
TOVP         3.036467
USG          5.441081

```

```

OWS          4.595000
DWS          4.242302
OBPM         9.587956
DBPM         4.853477
dtype: float64

```

```

[36]: #removing MP due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'TOV', 'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'OBPM', 'DBPM', ]
VIF(train,cols)

```

```

[36]: Season      1.136723
      Age         1.136248
      G          1.889153
      GS         3.298586
      X3PM       4.999364
      X3PP       2.025477
      X2PM      11.813506
      X2PP       4.034273
      EFG        5.215808
      FTA        9.197626
      FTP        1.577641
      ORB        9.793625
      DRB      11.681708
      AST       10.919832
      STL        5.349282
      BLK        5.235117
      TOV      14.067784
      PF         3.034813
      TS         5.863261
      X3PAR      4.485372
      FTR        3.077002
      ORBP       8.462237
      DRBP       5.655832
      ASTP       7.912304
      STLP       3.358746
      BLKP       5.026976
      TOVP       3.030833
      USG        4.892082
      OWS        4.472444
      DWS        4.173373
      OBPM       9.323727
      DBPM       4.748321

```

dtype: float64

```
[37]: #removing TOV due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PM', 'X2PP',
         'EFG', 'FTA', 'FTP', 'ORB', 'DRB', 'AST', 'STL', 'BLK',
         'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'OBPM', 'DBPM', ]
VIF(train,cols)
```

```
[37]: Season      1.135775
Age          1.130004
G            1.888278
GS           3.295046
X3PM         4.912694
X3PP         2.023405
X2PM        11.452405
X2PP         4.034139
EFG          5.207698
FTA          8.060751
FTP          1.574891
ORB          9.732625
DRB         11.674759
AST           7.014743
STL          5.347932
BLK          5.234317
PF           2.891596
TS           5.849500
X3PAR        4.485192
FTR          3.058400
ORBP         8.401629
DRBP         5.654377
ASTP         7.752107
STLP         3.358745
BLKP         5.017649
TOVP         2.392870
USG          4.619550
OWS          3.950771
DWS          4.103320
OBPM         9.231896
DBPM         4.719999
dtype: float64
```

```
[38]: #removing DRB due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PM', 'X2PP',
```

```

'EFG', 'FTA', 'FTP', 'ORB', 'AST', 'STL', 'BLK',
'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
'OBPM', 'DBPM', ]
VIF(train,cols)

```

```

[38]: Season      1.131626
      Age         1.105586
      G           1.886255
      GS          3.252811
      X3PM        4.682543
      X3PP        2.013417
      X2PM       10.554792
      X2PP        4.034062
      EFG         5.206740
      FTA         8.006049
      FTP         1.573864
      ORB         7.039333
      AST         7.014608
      STL         5.346409
      BLK         4.961322
      PF          2.889069
      TS          5.846426
      X3PAR       4.469559
      FTR         3.042172
      ORBP        7.391847
      DRBP        3.028741
      ASTP        7.719139
      STLP        3.356726
      BLKP        4.915779
      TOVP        2.392109
      USG         4.456246
      OWS         3.893081
      DWS         3.951947
      OBPM        9.223139
      DBPM        4.705253
      dtype: float64

```

```

[39]: #removing X2PM due to high VIF
      cols = [ 'Season', 'Age', 'G', 'GS',
               'X3PM', 'X3PP', 'X2PP',
               'EFG', 'FTA', 'FTP', 'ORB', 'AST', 'STL', 'BLK',
               'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
               'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
               'OBPM', 'DBPM', ]
      VIF(train,cols)

```

```
[39]: Season    1.130710
      Age      1.104392
      G        1.885560
      GS       3.044699
      X3PM     4.461442
      X3PP     1.996280
      X2PP     3.946231
      EFG      5.202824
      FTA      5.787522
      FTP      1.559904
      ORB      6.708103
      AST      6.563561
      STL      5.322994
      BLK      4.943520
      PF       2.874465
      TS       5.811265
      X3PAR    3.766829
      FTR      2.504369
      ORBP     7.205162
      DRBP     3.017469
      ASTP     7.228077
      STLP     3.338498
      BLKP     4.904424
      TOVP     2.389749
      USG      3.962929
      OWS      3.889772
      DWS      3.951828
      OBPM     8.998154
      DBPM     4.705084
      dtype: float64
```

```
[40]: #removing OBPM due to high VIF
      cols = [ 'Season', 'Age', 'G', 'GS',
                'X3PM', 'X3PP', 'X2PP',
                'EFG', 'FTA', 'FTP', 'ORB', 'AST', 'STL', 'BLK',
                'PF', 'TS', 'X3PAR', 'FTR', 'ORBP', 'DRBP',
                'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
                'DBPM', ]
      VIF(train,cols)
```

```
[40]: Season    1.100899
      Age      1.101521
      G        1.882005
      GS       3.044696
      X3PM     4.246389
      X3PP     1.993230
      X2PP     3.945502
```

```

EFG      5.063072
FTA      5.782452
FTP      1.543005
ORB      6.517056
AST      6.546866
STL      5.194886
BLK      4.859312
PF       2.545097
TS       3.049602
X3PAR    3.680976
FTR      2.487958
ORBP     6.846407
DRBP     2.964372
ASTP     5.738207
STLP     3.276234
BLKP     4.802364
TOVP     1.835915
USG      3.648232
OWS      3.490515
DWS      3.850792
DBPM     4.390282
dtype: float64

```

```

[41]: #removing ORBP due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
        'X3PM', 'X3PP', 'X2PP',
        'EFG', 'FTA', 'FTP', 'ORB', 'AST', 'STL', 'BLK',
        'PF', 'TS', 'X3PAR', 'FTR', 'DRBP',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        'DBPM', ]
VIF(train,cols)

```

```

[41]: Season    1.099357
      Age       1.101409
      G        1.879815
      GS       2.915875
      X3PM     4.153773
      X3PP     1.992038
      X2PP     3.945467
      EFG     5.062506
      FTA     5.580427
      FTP     1.542795
      ORB     4.394224
      AST     6.346549
      STL     5.190225
      BLK     4.488437
      PF      2.505498

```


TS	3.049306
X3PAR	3.169308
FTR	2.438070
DRBP	2.615965
ASTP	5.151984
STLP	3.221742
BLKP	4.156132
TOVP	1.730213
USG	3.633784
OWS	3.413857
DWS	3.839131
DBPM	4.159746

dtype: float64

```
[42]: #removing AST due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'EFG', 'FTA', 'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'X3PAR', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
VIF(train,cols)
```

```
[42]: Season    1.099079
      Age      1.098199
      G       1.879814
      GS      2.819469
      X3PM    4.140265
      X3PP    1.991777
      X2PP    3.945325
      EFG     5.060724
      FTA     5.276088
      FTP     1.540714
      ORB     4.225238
      STL     4.411035
      BLK     4.429971
      PF      2.500694
      TS      3.046562
      X3PAR   3.130705
      FTR     2.422135
      DRBP    2.585284
      ASTP    2.704411
      STLP    3.171858
      BLKP    4.104606
      TOVP    1.729317
      USG     3.476758
      OWS     3.409592
```

```
DWS      3.838223
DBPM     4.114434
dtype: float64
```

```
[43]: #removing FTA due to high VIF
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'EFG', 'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'X3PAR', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
VIF(train,cols)
```

```
[43]: Season      1.098972
Age          1.098197
G            1.878774
GS           2.729311
X3PM         4.134291
X3PP         1.990011
X2PP         3.931779
EFG          5.027074
FTP          1.535914
ORB          4.149917
STL          3.724103
BLK          4.199374
PF           2.499454
TS           2.996593
X3PAR        3.126218
FTR          1.714497
DRBP         2.582271
ASTP         2.683918
STLP         2.821396
BLKP         3.816001
TOVP         1.722360
USG          2.531905
OWS          3.116697
DWS          3.631480
DBPM         3.894991
dtype: float64
```

```
[153]: #removing EFG due to high VIF.
cols = [ 'Season', 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'X3PAR', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
```

```
VIF(train,cols)
```

```
[153]: Season    1.094741
      Age      1.093105
      G       2.036491
      GS      1.436180
      X3PM    3.778693
      X3PP    1.975416
      X2PP    1.932047
      FTP     1.430927
      ORB     3.720472
      STL     3.199244
      BLK     4.613826
      PF      2.526033
      TS      3.138067
      X3PAR   3.227589
      FTR     1.696305
      DRBP    2.379107
      ASTP    2.538724
      STLP    2.973322
      BLKP    4.366174
      TOVP    1.953010
      USG     2.311579
      OWS     2.740344
      DWS     3.031416
      DBPM    4.243942
      dtype: float64
```

```
[156]: #removed season b/c not relevant to predicting output
cols = [ 'Age', 'G', 'GS',
        'X3PM', 'X3PP', 'X2PP',
        'FTP', 'ORB', 'STL', 'BLK',
        'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:                MVP    R-squared:                0.022
```

```

Model: OLS Adj. R-squared: 0.010
Method: Least Squares F-statistic: 1.798
Date: Wed, 12 May 2021 Prob (F-statistic): 0.0114
Time: 19:24:14 Log-Likelihood: 16233.
No. Observations: 1892 AIC: -3.242e+04
Df Residuals: 1868 BIC: -3.228e+04
Df Model: 23
Covariance Type: nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	1.36e-05	1.69e-05	0.805	0.421	-1.95e-05	4.67e-05
Age	2.025e-07	2.58e-07	0.785	0.433	-3.04e-07	7.09e-07
G	-9.361e-08	8.3e-08	-1.128	0.259	-2.56e-07	6.91e-08
GS	1.318e-07	1.38e-07	0.956	0.339	-1.39e-07	4.02e-07
X3PM	2.804e-06	3.76e-06	0.745	0.456	-4.57e-06	1.02e-05
X3PP	-4.379e-06	9.06e-06	-0.484	0.629	-2.21e-05	1.34e-05
X2PP	1.086e-06	2.13e-05	0.051	0.959	-4.07e-05	4.29e-05
FTP	2.23e-07	8.84e-06	0.025	0.980	-1.71e-05	1.76e-05
ORB	1.576e-06	3.44e-06	0.458	0.647	-5.17e-06	8.32e-06
STL	8.593e-06	6.75e-06	1.273	0.203	-4.65e-06	2.18e-05
BLK	-3.613e-06	7.16e-06	-0.504	0.614	-1.77e-05	1.04e-05
PF	-2.587e-06	2.78e-06	-0.930	0.352	-8.04e-06	2.87e-06
TS	-5.772e-05	2.72e-05	-2.123	0.034	-0.000	-4.39e-06
FTR	7.147e-06	8.64e-06	0.827	0.408	-9.79e-06	2.41e-05
DRBP	-1.36e-07	2.85e-07	-0.477	0.633	-6.95e-07	4.23e-07
X3PAR	2.589e-06	8.59e-06	0.301	0.763	-1.43e-05	1.94e-05
ASTP	-5.788e-08	2.09e-07	-0.277	0.782	-4.68e-07	3.52e-07
STLP	-3.083e-06	2.57e-06	-1.197	0.231	-8.13e-06	1.97e-06
BLKP	1.336e-07	1.41e-06	0.095	0.924	-2.63e-06	2.9e-06
TOVP	3.205e-07	2.85e-07	1.126	0.260	-2.38e-07	8.79e-07
USG	6.901e-07	3.66e-07	1.886	0.059	-2.77e-08	1.41e-06
OWS	5.188e-06	1.39e-06	3.744	0.000	2.47e-06	7.91e-06
DWS	-9.126e-07	2.39e-06	-0.382	0.703	-5.6e-06	3.78e-06
DBPM	3.445e-06	1.63e-06	2.113	0.035	2.48e-07	6.64e-06
=====						
Omnibus:		5681.464	Durbin-Watson:		2.010	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		257689789.971	
Skew:		42.052	Prob(JB):		0.00	
Kurtosis:		1809.023	Cond. No.		2.06e+03	
=====						

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.06e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[46]: #removed X3PAR b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MVP    R-squared:                0.201
Model:                  OLS    Adj. R-squared:           0.195
Method:                 Least Squares    F-statistic:         36.67
Date:                  Wed, 12 May 2021    Prob (F-statistic):    1.94e-138
Time:                  19:19:01    Log-Likelihood:        4695.5
No. Observations:      3231    AIC:                  -9345.
Df Residuals:          3208    BIC:                  -9205.
Df Model:              22
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0595	0.017	3.599	0.000	0.027	0.092
Age	-0.0004	0.000	-1.458	0.145	-0.001	0.000
G	-6.818e-05	7.73e-05	-0.882	0.378	-0.000	8.34e-05
GS	-0.0001	5.63e-05	-2.318	0.021	-0.000	-2.01e-05
X3PM	0.0115	0.002	4.836	0.000	0.007	0.016
X3PP	-0.0169	0.009	-1.972	0.049	-0.034	-9.88e-05
X2PP	0.0716	0.022	3.257	0.001	0.029	0.115
FTP	-0.0147	0.009	-1.552	0.121	-0.033	0.004
ORB	-0.0031	0.002	-1.278	0.201	-0.008	0.002
STL	0.0196	0.005	4.300	0.000	0.011	0.028
BLK	0.0135	0.004	3.081	0.002	0.005	0.022
PF	-0.0145	0.002	-6.249	0.000	-0.019	-0.010
TS	-0.1977	0.024	-8.218	0.000	-0.245	-0.151
FTR	0.0123	0.008	1.462	0.144	-0.004	0.029
DRBP	0.0002	0.000	0.885	0.376	-0.000	0.001
ASTP	-0.0005	0.000	-2.684	0.007	-0.001	-0.000
STLP	-0.0120	0.002	-4.966	0.000	-0.017	-0.007

BLKP	-0.0043	0.001	-3.757	0.000	-0.007	-0.002
TOVP	0.0016	0.000	5.934	0.000	0.001	0.002
USG	0.0023	0.000	7.459	0.000	0.002	0.003
OWS	0.0130	0.001	15.363	0.000	0.011	0.015
DWS	-0.0044	0.002	-2.620	0.009	-0.008	-0.001
DBPM	0.0112	0.001	7.580	0.000	0.008	0.014

```
=====
Omnibus:                4869.946    Durbin-Watson:                1.244
Prob(Omnibus):           0.000    Jarque-Bera (JB):            1899462.383
Skew:                    9.285    Prob(JB):                     0.00
Kurtosis:                120.322    Cond. No.                    2.32e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.32e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[47]: #removed 'G' games b/c not statistically significant
cols = [ 'Age', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MVP    R-squared:                0.201
Model:                  OLS    Adj. R-squared:           0.196
Method:                 Least Squares    F-statistic:              38.38
Date:                   Wed, 12 May 2021    Prob (F-statistic):       4.51e-139
Time:                   19:19:02    Log-Likelihood:           4695.1
No. Observations:       3231    AIC:                      -9346.
Df Residuals:           3209    BIC:                      -9212.
Df Model:               21
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	0.0574	0.016	3.509	0.000	0.025	0.090
Age	-0.0003	0.000	-1.398	0.162	-0.001	0.000
GS	-0.0001	5.5e-05	-2.555	0.011	-0.000	-3.27e-05
X3PM	0.0114	0.002	4.792	0.000	0.007	0.016
X3PP	-0.0173	0.009	-2.022	0.043	-0.034	-0.001
X2PP	0.0693	0.022	3.176	0.002	0.027	0.112
FTP	-0.0156	0.009	-1.656	0.098	-0.034	0.003
ORB	-0.0032	0.002	-1.294	0.196	-0.008	0.002
STL	0.0198	0.005	4.365	0.000	0.011	0.029
BLK	0.0139	0.004	3.165	0.002	0.005	0.022
PF	-0.0148	0.002	-6.481	0.000	-0.019	-0.010
TS	-0.1981	0.024	-8.239	0.000	-0.245	-0.151
FTR	0.0126	0.008	1.501	0.133	-0.004	0.029
DRBP	0.0002	0.000	0.969	0.333	-0.000	0.001
ASTP	-0.0005	0.000	-2.722	0.007	-0.001	-0.000
STLP	-0.0121	0.002	-4.993	0.000	-0.017	-0.007
BLKP	-0.0044	0.001	-3.809	0.000	-0.007	-0.002
TOVP	0.0016	0.000	5.999	0.000	0.001	0.002
USG	0.0023	0.000	7.682	0.000	0.002	0.003
OWS	0.0130	0.001	15.389	0.000	0.011	0.015
DWS	-0.0049	0.002	-3.134	0.002	-0.008	-0.002
DBPM	0.0114	0.001	7.763	0.000	0.008	0.014

```
=====
Omnibus:                4869.595    Durbin-Watson:                1.244
Prob(Omnibus):           0.000    Jarque-Bera (JB):            1898225.383
Skew:                    9.284    Prob(JB):                     0.00
Kurtosis:                120.283    Cond. No.                    1.54e+03
=====
```

Warnings:

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

[2] The condition number is large, 1.54e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[48]: #removed 'DRBP' games b/c not statistically significant
cols = [ 'Age', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MVP      R-squared:          0.201
Model:                  OLS      Adj. R-squared:       0.196
Method:                 Least Squares  F-statistic:       40.25
Date:                  Wed, 12 May 2021  Prob (F-statistic):  1.11e-139
Time:                  19:19:02   Log-Likelihood:     4694.6
No. Observations:      3231      AIC:               -9347.
Df Residuals:          3210      BIC:               -9220.
Df Model:               20
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0600	0.016	3.711	0.000	0.028	0.092
Age	-0.0003	0.000	-1.347	0.178	-0.001	0.000
GS	-0.0001	5.49e-05	-2.622	0.009	-0.000	-3.63e-05
X3PM	0.0114	0.002	4.777	0.000	0.007	0.016
X3PP	-0.0172	0.009	-2.012	0.044	-0.034	-0.000
X2PP	0.0697	0.022	3.195	0.001	0.027	0.113
FTP	-0.0161	0.009	-1.707	0.088	-0.035	0.002
ORB	-0.0023	0.002	-1.018	0.309	-0.007	0.002
STL	0.0198	0.005	4.366	0.000	0.011	0.029
BLK	0.0130	0.004	3.032	0.002	0.005	0.021
PF	-0.0148	0.002	-6.482	0.000	-0.019	-0.010
TS	-0.2000	0.024	-8.343	0.000	-0.247	-0.153
FTR	0.0129	0.008	1.539	0.124	-0.004	0.029
ASTP	-0.0005	0.000	-3.114	0.002	-0.001	-0.000
STLP	-0.0125	0.002	-5.287	0.000	-0.017	-0.008
BLKP	-0.0041	0.001	-3.686	0.000	-0.006	-0.002
TOVP	0.0016	0.000	6.362	0.000	0.001	0.002
USG	0.0024	0.000	8.186	0.000	0.002	0.003
OWS	0.0130	0.001	15.402	0.000	0.011	0.015
DWS	-0.0047	0.002	-3.033	0.002	-0.008	-0.002
DBPM	0.0117	0.001	8.220	0.000	0.009	0.014

```
=====
Omnibus:                4871.045   Durbin-Watson:          1.244
Prob(Omnibus):           0.000     Jarque-Bera (JB):       1898740.151
Skew:                    9.290      Prob(JB):               0.00
Kurtosis:                120.298    Cond. No.               1.48e+03
=====
```


Warnings:

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

[2] The condition number is large, 1.48e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[49]: #removed 'ORB' games b/c not statistically significant
cols = [ 'Age', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MVP      R-squared:                0.200
Model:                  OLS      Adj. R-squared:            0.196
Method:                 Least Squares      F-statistic:         42.32
Date:                   Wed, 12 May 2021    Prob (F-statistic):      2.79e-140
Time:                   19:19:02           Log-Likelihood:       4694.1
No. Observations:       3231             AIC:                  -9348.
Df Residuals:           3211             BIC:                  -9227.
Df Model:                19
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0592	0.016	3.668	0.000	0.028	0.091
Age	-0.0003	0.000	-1.336	0.182	-0.001	0.000
GS	-0.0002	5.44e-05	-2.787	0.005	-0.000	-4.5e-05
X3PM	0.0122	0.002	5.535	0.000	0.008	0.017
X3PP	-0.0159	0.008	-1.879	0.060	-0.032	0.001
X2PP	0.0682	0.022	3.132	0.002	0.026	0.111
FTP	-0.0153	0.009	-1.626	0.104	-0.034	0.003
STL	0.0197	0.005	4.337	0.000	0.011	0.029
BLK	0.0115	0.004	2.856	0.004	0.004	0.019
PF	-0.0156	0.002	-7.227	0.000	-0.020	-0.011
TS	-0.1999	0.024	-8.338	0.000	-0.247	-0.153
FTR	0.0129	0.008	1.543	0.123	-0.003	0.029

ASTP	-0.0005	0.000	-2.973	0.003	-0.001	-0.000
STLP	-0.0126	0.002	-5.311	0.000	-0.017	-0.008
BLKP	-0.0040	0.001	-3.612	0.000	-0.006	-0.002
TOVP	0.0016	0.000	6.362	0.000	0.001	0.002
USG	0.0024	0.000	8.164	0.000	0.002	0.003
OWS	0.0128	0.001	15.502	0.000	0.011	0.014
DWS	-0.0048	0.002	-3.126	0.002	-0.008	-0.002
DBPM	0.0119	0.001	8.436	0.000	0.009	0.015

```
=====
Omnibus:                4869.255    Durbin-Watson:                1.241
Prob(Omnibus):           0.000    Jarque-Bera (JB):            1892741.851
Skew:                    9.284    Prob(JB):                     0.00
Kurtosis:               120.109    Cond. No.                     1.48e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.48e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[50]: #removed season b/c not relevant to predicting output
cols = [ 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['MVP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MVP    R-squared:                0.200
Model:                  OLS    Adj. R-squared:           0.195
Method:                 Least Squares    F-statistic:              44.56
Date:                   Wed, 12 May 2021    Prob (F-statistic):       9.84e-141
Time:                   19:19:02    Log-Likelihood:           4693.2
No. Observations:       3231    AIC:                      -9348.
Df Residuals:           3212    BIC:                      -9233.
Df Model:                18
Covariance Type:        nonrobust
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0491	0.014	3.443	0.001	0.021	0.077
GS	-0.0002	5.44e-05	-2.806	0.005	-0.000	-4.6e-05
X3PM	0.0120	0.002	5.455	0.000	0.008	0.016
X3PP	-0.0156	0.008	-1.846	0.065	-0.032	0.001
X2PP	0.0683	0.022	3.134	0.002	0.026	0.111
FTP	-0.0157	0.009	-1.676	0.094	-0.034	0.003
STL	0.0201	0.005	4.443	0.000	0.011	0.029
BLK	0.0117	0.004	2.895	0.004	0.004	0.020
PF	-0.0156	0.002	-7.242	0.000	-0.020	-0.011
TS	-0.1986	0.024	-8.290	0.000	-0.246	-0.152
FTR	0.0134	0.008	1.597	0.110	-0.003	0.030
ASTP	-0.0005	0.000	-3.051	0.002	-0.001	-0.000
STLP	-0.0124	0.002	-5.236	0.000	-0.017	-0.008
BLKP	-0.0040	0.001	-3.557	0.000	-0.006	-0.002
TOVP	0.0016	0.000	6.345	0.000	0.001	0.002
USG	0.0024	0.000	8.306	0.000	0.002	0.003
OWS	0.0128	0.001	15.463	0.000	0.011	0.014
DWS	-0.0049	0.002	-3.174	0.002	-0.008	-0.002
DBPM	0.0117	0.001	8.338	0.000	0.009	0.014
Omnibus:		4873.501	Durbin-Watson:		1.240	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		1901763.462	
Skew:		9.299	Prob(JB):		0.00	
Kurtosis:		120.390	Cond. No.		1.33e+03	

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.33e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[51]: import statsmodels.formula.api as smf

MVPLinear = smf.ols(formula='MVP ~ GS + X3PM + X3PP + X2PP + STL+ BLK + PF + TS_
↪+ ASTP + STLP + BLKP + TOVP + USG + OWS + DWS + DBPM',
                    data=train).fit()
print(MVPLinear.summary())
```

OLS Regression Results

Dep. Variable:	MVP	R-squared:	0.198
Model:	OLS	Adj. R-squared:	0.194
Method:	Least Squares	F-statistic:	49.71
Date:	Wed, 12 May 2021	Prob (F-statistic):	3.33e-141

Time: 19:19:02 Log-Likelihood: 4690.3
 No. Observations: 3231 AIC: -9347.
 Df Residuals: 3214 BIC: -9243.
 Df Model: 16
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.0394	0.013	2.946	0.003	0.013	0.066
GS	-0.0002	5.44e-05	-2.904	0.004	-0.000	-5.13e-05
X3PM	0.0109	0.002	5.057	0.000	0.007	0.015
X3PP	-0.0192	0.008	-2.311	0.021	-0.035	-0.003
X2PP	0.0706	0.022	3.284	0.001	0.028	0.113
STL	0.0206	0.005	4.551	0.000	0.012	0.029
BLK	0.0113	0.004	2.810	0.005	0.003	0.019
PF	-0.0156	0.002	-7.235	0.000	-0.020	-0.011
TS	-0.1992	0.023	-8.678	0.000	-0.244	-0.154
ASTP	-0.0006	0.000	-3.502	0.000	-0.001	-0.000
STLP	-0.0118	0.002	-5.014	0.000	-0.016	-0.007
BLKP	-0.0035	0.001	-3.191	0.001	-0.006	-0.001
TOVP	0.0018	0.000	7.153	0.000	0.001	0.002
USG	0.0024	0.000	8.322	0.000	0.002	0.003
OWS	0.0131	0.001	16.372	0.000	0.012	0.015
DWS	-0.0048	0.002	-3.131	0.002	-0.008	-0.002
DBPM	0.0114	0.001	8.184	0.000	0.009	0.014
Omnibus:	4877.428		Durbin-Watson:	1.241		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	1907434.996		
Skew:	9.314		Prob(JB):	0.00		
Kurtosis:	120.565		Cond. No.	1.28e+03		

Warnings:

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

[2] The condition number is large, 1.28e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[52]: # compute out-of-sample R-squared using the test set
def OSR2(model, df_train, df_test, dependent_var):
    y_test = df_test[dependent_var]
    y_pred = model.predict(df_test)
    SSE = np.sum((y_test - y_pred)**2)
    SST = np.sum((y_test - np.mean(df_train[dependent_var]))**2)
    return 1 - SSE/SST
```

```
[53]: OSR2(MVPLinear, train, test, 'MVP')
```

```
[53]: 0.22559973569316927
```

```
[54]: dataset = pd.read_csv('mip_stats.csv')
train = dataset[dataset['Season'] <= 2017]
test = dataset[dataset['Season'] > 2017]
test_2010 = test = dataset[dataset['Season'] > 2010]
```

```
[55]: dataset
```

```
[55]:
```

	Player	Season	G	GS	MP	FGM	FGA	FGP	X3PM	\		
0	A.J. Price	2011.0	-6.0	-2.0	0.5	-0.3	0.1	-0.054	-0.3			
1	A.J. Price	2012.0	-6.0	1.0	-3.0	-1.0	-2.4	-0.017	-0.2			
2	A.J. Price	2013.0	13.0	21.0	9.5	1.5	3.2	0.051	0.6			
3	A.J. Price	2014.0	-29.0	-22.0	-18.9	-2.1	-5.6	0.023	-1.0			
4	A.J. Price	2015.0	-2.0	0.0	9.0	1.3	3.7	-0.041	0.4			
...			
3405	Zaza Pachulia	2016.0	3.0	24.0	2.7	-0.4	-1.0	0.012	0.0			
3406	Zaza Pachulia	2017.0	-6.0	1.0	-8.3	-0.6	-1.8	0.068	0.0			
3407	Zaza Pachulia	2018.0	-1.0	-13.0	-4.0	-0.1	-0.6	0.030	0.0			
3408	Zaza Pachulia	2019.0	-1.0	-54.0	-1.2	-0.9	-1.0	-0.124	0.0			
3409	Zydrunas Ilgauskas	2011.0	8.0	45.0	-5.0	-0.7	-2.4	0.065	-0.2			
	X3PA	...	USG	OWS	DWS	WS	WS48	OBPM	DBPM	BPM	VORP	MIP
0	-0.1	...	0.0	-0.8	-0.1	-0.9	-0.045	-1.8	-0.7	-2.4	-0.5	0.0
1	-1.0	...	-5.0	0.6	-0.2	0.4	0.043	1.3	0.6	1.9	0.3	0.0
2	1.5	...	0.3	0.8	0.7	1.5	0.021	0.4	0.1	0.4	0.4	0.0
3	-2.7	...	5.2	-1.1	-1.2	-2.2	-0.096	-2.2	-2.0	-4.2	-0.6	0.0
4	1.4	...	-5.0	0.9	0.5	1.4	0.057	1.5	1.3	2.8	0.1	0.0
...
3405	0.0	...	-10.1	-1.3	0.5	-1.0	-0.014	-2.1	0.8	-1.4	-0.7	0.0
3406	0.0	...	-0.9	-1.1	-0.2	-1.3	0.033	-0.8	1.5	0.8	-0.1	0.0
3407	0.0	...	1.7	-0.5	-1.1	-1.4	-0.017	0.4	-0.9	-0.6	-0.3	0.0
3408	0.1	...	-2.0	-0.7	-0.1	-1.0	-0.033	-2.1	0.9	-1.2	-0.3	0.0
3409	-0.4	...	-3.0	0.5	-0.1	0.4	0.034	0.0	1.1	1.1	0.4	0.0

```
[3410 rows x 48 columns]
```

```
[157]: #use this to test award winner for each year and award
datasetmip = pd.read_csv('smoydata142.csv')
trainmip = datasetmip[datasetmip['Season'] <= 2017]
testmip = datasetmip[datasetmip['Season'] > 2017]
data2009 = datasetmip[datasetmip['Season'] == 2020.0]
testt = dataset[dataset['Season'] == 2020]
graphh = testt[testt['MIP']>0]
funn = graphh[['Player', 'MIP']].sort_values(by='MIP',ascending = False)
funn
```

```
[157]:
```

	Player	MIP
450	Brandon Ingram	0.652
334	Bam Adebayo	0.590
1971	Jayson Tatum	0.114
1123	Devonte' Graham	0.100
3813	Shai Gilgeous-Alexander	0.042
3344	Pascal Siakam	0.026
705	Christian Wood	0.022
4155	Trae Young	0.020
1420	Fred VanVleet	0.012
1968	Jaylen Brown	0.006
2915	Markelle Fultz	0.004
1221	Duncan Robinson	0.002
3889	Spencer Dinwiddie	0.002

Sixth Man of the Year (SMOY)

```
[158]: datasett = pd.read_csv('smoydata142.csv')
train = datasett[datasett['Season'] <= 2017]
test = datasett[datasett['Season'] > 2017]

y_train = train['MVP']
y_test = test['MVP']
```

```
[159]: #removed season b/c not relevant to predicting the output variable
cols = [ 'Age', 'G', 'GS',
        'X3PM', 'X3PP', 'X2PP',
        'FTP', 'ORB', 'STL', 'BLK',
        'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
        'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        'DBPM', ]

dataset = pd.read_csv('update1422.csv')
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY    R-squared:                0.149
Model:                  OLS    Adj. R-squared:           0.139
Method:                 Least Squares    F-statistic:           14.24
Date:                   Wed, 12 May 2021    Prob (F-statistic):      8.97e-51
```

Time: 19:28:00 Log-Likelihood: 2684.8
 No. Observations: 1892 AIC: -5322.
 Df Residuals: 1868 BIC: -5188.
 Df Model: 23
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0446	0.022	2.050	0.041	0.002	0.087
Age	0.0005	0.000	1.521	0.128	-0.000	0.001
G	-5.714e-05	0.000	-0.535	0.593	-0.000	0.000
GS	-0.0009	0.000	-5.050	0.000	-0.001	-0.001
X3PM	0.0332	0.005	6.857	0.000	0.024	0.043
X3PP	-0.0152	0.012	-1.302	0.193	-0.038	0.008
X2PP	0.0644	0.027	2.349	0.019	0.011	0.118
FTP	-0.0052	0.011	-0.458	0.647	-0.028	0.017
ORB	0.0023	0.004	0.521	0.603	-0.006	0.011
STL	0.0271	0.009	3.115	0.002	0.010	0.044
BLK	0.0182	0.009	1.976	0.048	0.000	0.036
PF	-0.0115	0.004	-3.205	0.001	-0.019	-0.004
TS	-0.1642	0.035	-4.691	0.000	-0.233	-0.096
FTR	0.0030	0.011	0.274	0.784	-0.019	0.025
DRBP	-0.0008	0.000	-2.137	0.033	-0.002	-6.43e-05
X3PAR	-0.0407	0.011	-3.676	0.000	-0.062	-0.019
ASTP	-0.0008	0.000	-3.042	0.002	-0.001	-0.000
STLP	-0.0115	0.003	-3.457	0.001	-0.018	-0.005
BLKP	-0.0045	0.002	-2.497	0.013	-0.008	-0.001
TOVP	0.0014	0.000	3.844	0.000	0.001	0.002
USG	0.0013	0.000	2.826	0.005	0.000	0.002
OWS	0.0136	0.002	7.608	0.000	0.010	0.017
DWS	0.0046	0.003	1.479	0.139	-0.001	0.011
DBPM	0.0050	0.002	2.368	0.018	0.001	0.009

Omnibus: 2876.639 Durbin-Watson: 1.769
 Prob(Omnibus): 0.000 Jarque-Bera (JB): 960973.714
 Skew: 9.353 Prob(JB): 0.00
 Kurtosis: 111.812 Cond. No. 2.06e+03

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.06e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[61]: *#removed FTR b/c not statistically significant*

```

cols = [ 'Age', 'G', 'GS',
          'X3PM', 'X3PP', 'X2PP',
          'FTP', 'ORB', 'STL', 'BLK',
          'PF', 'TS', 'DRBP', 'X3PAR',
          'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
          'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          SMOY      R-squared:                0.149
Model:                  OLS      Adj. R-squared:           0.139
Method:                 Least Squares      F-statistic:        14.89
Date:                   Wed, 12 May 2021    Prob (F-statistic):    2.37e-51
Time:                   19:19:03    Log-Likelihood:       2684.7
No. Observations:      1892      AIC:                 -5323.
Df Residuals:          1869      BIC:                 -5196.
Df Model:               22
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0450	0.022	2.071	0.038	0.002	0.088
Age	0.0005	0.000	1.511	0.131	-0.000	0.001
G	-5.78e-05	0.000	-0.541	0.588	-0.000	0.000
GS	-0.0009	0.000	-5.070	0.000	-0.001	-0.001
X3PM	0.0332	0.005	6.855	0.000	0.024	0.043
X3PP	-0.0154	0.012	-1.325	0.185	-0.038	0.007
X2PP	0.0638	0.027	2.335	0.020	0.010	0.117
FTP	-0.0054	0.011	-0.476	0.634	-0.028	0.017
ORB	0.0023	0.004	0.513	0.608	-0.006	0.011
STL	0.0271	0.009	3.116	0.002	0.010	0.044
BLK	0.0181	0.009	1.967	0.049	5.11e-05	0.036
PF	-0.0115	0.004	-3.199	0.001	-0.018	-0.004
TS	-0.1625	0.034	-4.723	0.000	-0.230	-0.095
DRBP	-0.0008	0.000	-2.141	0.032	-0.002	-6.59e-05
X3PAR	-0.0415	0.011	-3.908	0.000	-0.062	-0.021
ASTP	-0.0008	0.000	-3.125	0.002	-0.001	-0.000
STLP	-0.0114	0.003	-3.449	0.001	-0.018	-0.005
BLKP	-0.0045	0.002	-2.484	0.013	-0.008	-0.001

TOVP	0.0014	0.000	3.985	0.000	0.001	0.002
USG	0.0013	0.000	2.824	0.005	0.000	0.002
OWS	0.0137	0.002	7.755	0.000	0.010	0.017
DWS	0.0046	0.003	1.478	0.140	-0.001	0.011
DBPM	0.0049	0.002	2.353	0.019	0.001	0.009

Omnibus:	2876.791	Durbin-Watson:	1.769
Prob(Omnibus):	0.000	Jarque-Bera (JB):	961369.871
Skew:	9.354	Prob(JB):	0.00
Kurtosis:	111.835	Cond. No.	2.03e+03

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.03e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[62]: #removed X3PP b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	SMOY	R-squared:	0.148
Model:	OLS	Adj. R-squared:	0.139
Method:	Least Squares	F-statistic:	15.51
Date:	Wed, 12 May 2021	Prob (F-statistic):	1.35e-51
Time:	19:19:03	Log-Likelihood:	2683.9
No. Observations:	1892	AIC:	-5324.
Df Residuals:	1870	BIC:	-5202.
Df Model:	21		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	0.0439	0.022	2.022	0.043	0.001	0.086
Age	0.0005	0.000	1.563	0.118	-0.000	0.001
G	-6.454e-05	0.000	-0.605	0.545	-0.000	0.000
GS	-0.0009	0.000	-5.103	0.000	-0.001	-0.001
X3PM	0.0319	0.005	6.725	0.000	0.023	0.041
X2PP	0.0678	0.027	2.493	0.013	0.014	0.121
FTP	-0.0067	0.011	-0.593	0.553	-0.029	0.015
ORB	0.0030	0.004	0.689	0.491	-0.006	0.012
STL	0.0267	0.009	3.073	0.002	0.010	0.044
BLK	0.0182	0.009	1.981	0.048	0.000	0.036
PF	-0.0111	0.004	-3.110	0.002	-0.018	-0.004
TS	-0.1684	0.034	-4.937	0.000	-0.235	-0.101
DRBP	-0.0008	0.000	-2.177	0.030	-0.002	-7.92e-05
X3PAR	-0.0435	0.011	-4.135	0.000	-0.064	-0.023
ASTP	-0.0008	0.000	-3.196	0.001	-0.001	-0.000
STLP	-0.0116	0.003	-3.542	0.000	-0.018	-0.005
BLKP	-0.0044	0.002	-2.458	0.014	-0.008	-0.001
TOVP	0.0015	0.000	4.077	0.000	0.001	0.002
USG	0.0013	0.000	2.831	0.005	0.000	0.002
OWS	0.0137	0.002	7.789	0.000	0.010	0.017
DWS	0.0044	0.003	1.424	0.155	-0.002	0.010
DBPM	0.0051	0.002	2.453	0.014	0.001	0.009

```
=====
Omnibus:                2878.350    Durbin-Watson:                1.767
Prob(Omnibus):           0.000    Jarque-Bera (JB):            963345.056
Skew:                    9.364    Prob(JB):                     0.00
Kurtosis:               111.946    Cond. No.                    2.01e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.01e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[63]: #removed FTP b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY      R-squared:                0.148
Model:                  OLS      Adj. R-squared:           0.139
Method:                 Least Squares      F-statistic:         16.28
Date:                   Wed, 12 May 2021    Prob (F-statistic):      3.89e-52
Time:                   19:19:03    Log-Likelihood:         2683.7
No. Observations:      1892      AIC:                   -5325.
Df Residuals:          1871      BIC:                   -5209.
Df Model:               20
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0411	0.021	1.940	0.053	-0.000	0.083
Age	0.0005	0.000	1.546	0.122	-0.000	0.001
G	-6.933e-05	0.000	-0.652	0.515	-0.000	0.000
GS	-0.0009	0.000	-5.109	0.000	-0.001	-0.001
X3PM	0.0317	0.005	6.701	0.000	0.022	0.041
X2PP	0.0698	0.027	2.592	0.010	0.017	0.123
ORB	0.0032	0.004	0.739	0.460	-0.005	0.012
STL	0.0266	0.009	3.065	0.002	0.010	0.044
BLK	0.0184	0.009	1.994	0.046	0.000	0.036
PF	-0.0112	0.004	-3.141	0.002	-0.018	-0.004
TS	-0.1728	0.033	-5.193	0.000	-0.238	-0.108
DRBP	-0.0008	0.000	-2.157	0.031	-0.002	-7.17e-05
X3PAR	-0.0435	0.011	-4.138	0.000	-0.064	-0.023
ASTP	-0.0009	0.000	-3.240	0.001	-0.001	-0.000
STLP	-0.0116	0.003	-3.521	0.000	-0.018	-0.005
BLKP	-0.0044	0.002	-2.438	0.015	-0.008	-0.001
TOVP	0.0015	0.000	4.140	0.000	0.001	0.002
USG	0.0013	0.000	2.778	0.006	0.000	0.002
OWS	0.0137	0.002	7.810	0.000	0.010	0.017
DWS	0.0043	0.003	1.414	0.158	-0.002	0.010
DBPM	0.0051	0.002	2.458	0.014	0.001	0.009

```
=====
Omnibus:                2878.889      Durbin-Watson:           1.767
Prob(Omnibus):          0.000      Jarque-Bera (JB):       964330.897
Skew:                   9.367      Prob(JB):               0.00
Kurtosis:               112.002      Cond. No.:               1.98e+03
=====
```

Warnings:

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

[2] The condition number is large, 1.98e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[64]: #removed DBPM b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',  ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY    R-squared:                0.145
Model:                  OLS    Adj. R-squared:           0.137
Method:                 Least Squares    F-statistic:        16.77
Date:                  Wed, 12 May 2021    Prob (F-statistic):    1.61e-51
Time:                  19:19:03    Log-Likelihood:       2680.6
No. Observations:      1892    AIC:                -5321.
Df Residuals:          1872    BIC:                -5210.
Df Model:              19
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0233	0.020	1.168	0.243	-0.016	0.062
Age	0.0006	0.000	1.666	0.096	-9.81e-05	0.001
G	-9.612e-05	0.000	-0.907	0.364	-0.000	0.000
GS	-0.0009	0.000	-5.337	0.000	-0.001	-0.001
X3PM	0.0305	0.005	6.474	0.000	0.021	0.040
X2PP	0.0622	0.027	2.321	0.020	0.010	0.115
ORB	0.0009	0.004	0.208	0.836	-0.007	0.009
STL	0.0246	0.009	2.846	0.004	0.008	0.042
BLK	0.0209	0.009	2.282	0.023	0.003	0.039
PF	-0.0107	0.004	-2.991	0.003	-0.018	-0.004
TS	-0.1350	0.030	-4.569	0.000	-0.193	-0.077
DRBP	-0.0006	0.000	-1.602	0.109	-0.001	0.000
X3PAR	-0.0423	0.011	-4.019	0.000	-0.063	-0.022
ASTP	-0.0007	0.000	-2.738	0.006	-0.001	-0.000

STLP	-0.0067	0.003	-2.552	0.011	-0.012	-0.002
BLKP	-0.0036	0.002	-2.031	0.042	-0.007	-0.000
TOVP	0.0013	0.000	3.795	0.000	0.001	0.002
USG	0.0007	0.000	1.696	0.090	-0.000	0.001
OWS	0.0129	0.002	7.456	0.000	0.009	0.016
DWS	0.0079	0.003	2.884	0.004	0.003	0.013

```
=====
Omnibus:                2886.709    Durbin-Watson:                1.770
Prob(Omnibus):           0.000    Jarque-Bera (JB):            978137.436
Skew:                    9.414    Prob(JB):                     0.00
Kurtosis:                112.787    Cond. No.                    1.80e+03
=====
```

Warnings:

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

[2] The condition number is large, 1.8e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[65]: #removed USG b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'OWS', 'DWS', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY    R-squared:                0.144
Model:                  OLS     Adj. R-squared:          0.136
Method:                 Least Squares    F-statistic:            17.53
Date:                   Wed, 12 May 2021    Prob (F-statistic):      1.48e-51
Time:                   19:19:03    Log-Likelihood:          2679.2
No. Observations:       1892    AIC:                     -5320.
Df Residuals:           1873    BIC:                     -5215.
Df Model:                18
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]

const	0.0394	0.018	2.253	0.024	0.005	0.074
Age	0.0005	0.000	1.385	0.166	-0.000	0.001
G	-0.0001	0.000	-1.001	0.317	-0.000	0.000
GS	-0.0010	0.000	-5.437	0.000	-0.001	-0.001
X3PM	0.0330	0.004	7.386	0.000	0.024	0.042
X2PP	0.0624	0.027	2.326	0.020	0.010	0.115
ORB	0.0015	0.004	0.360	0.719	-0.007	0.010
STL	0.0226	0.009	2.631	0.009	0.006	0.039
BLK	0.0212	0.009	2.314	0.021	0.003	0.039
PF	-0.0104	0.004	-2.917	0.004	-0.017	-0.003
TS	-0.1374	0.030	-4.653	0.000	-0.195	-0.079
DRBP	-0.0005	0.000	-1.521	0.128	-0.001	0.000
X3PAR	-0.0475	0.010	-4.717	0.000	-0.067	-0.028
ASTP	-0.0005	0.000	-2.236	0.025	-0.001	-6.46e-05
STLP	-0.0061	0.003	-2.357	0.019	-0.011	-0.001
BLKP	-0.0038	0.002	-2.169	0.030	-0.007	-0.000
TOVP	0.0011	0.000	3.417	0.001	0.000	0.002
OWS	0.0127	0.002	7.384	0.000	0.009	0.016
DWS	0.0080	0.003	2.934	0.003	0.003	0.013

```
=====
Omnibus:                2886.633    Durbin-Watson:                1.767
Prob(Omnibus):           0.000    Jarque-Bera (JB):            976727.636
Skew:                    9.415    Prob(JB):                     0.00
Kurtosis:                112.705    Cond. No.                     1.74e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.74e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[66]: #removed TOVP b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'OWS', 'DWS', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```

=====
Dep. Variable:          SMOY      R-squared:          0.139
Model:                  OLS       Adj. R-squared:       0.131
Method:                 Least Squares   F-statistic:        17.77
Date:                   Wed, 12 May 2021   Prob (F-statistic):  8.57e-50
Time:                   19:19:03    Log-Likelihood:     2673.3
No. Observations:      1892      AIC:                -5311.
Df Residuals:          1874      BIC:                -5211.
Df Model:               17
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0430	0.018	2.455	0.014	0.009	0.077
Age	0.0005	0.000	1.608	0.108	-0.000	0.001
G	-0.0001	0.000	-1.215	0.225	-0.000	7.91e-05
GS	-0.0009	0.000	-5.231	0.000	-0.001	-0.001
X3PM	0.0316	0.004	7.076	0.000	0.023	0.040
X2PP	0.0619	0.027	2.302	0.021	0.009	0.115
ORB	0.0022	0.004	0.508	0.611	-0.006	0.011
STL	0.0205	0.009	2.388	0.017	0.004	0.037
BLK	0.0168	0.009	1.843	0.065	-0.001	0.035
PF	-0.0085	0.004	-2.412	0.016	-0.015	-0.002
TS	-0.1254	0.029	-4.263	0.000	-0.183	-0.068
DRBP	-0.0006	0.000	-1.570	0.117	-0.001	0.000
X3PAR	-0.0505	0.010	-5.018	0.000	-0.070	-0.031
ASTP	-0.0002	0.000	-0.765	0.444	-0.001	0.000
STLP	-0.0055	0.003	-2.100	0.036	-0.011	-0.000
BLKP	-0.0027	0.002	-1.570	0.117	-0.006	0.001
OWS	0.0110	0.002	6.657	0.000	0.008	0.014
DWS	0.0083	0.003	3.034	0.002	0.003	0.014

```

=====
Omnibus:                2899.616   Durbin-Watson:          1.754
Prob(Omnibus):           0.000     Jarque-Bera (JB):       1000081.384
Skew:                    9.493      Prob(JB):               0.00
Kurtosis:                114.020    Cond. No.               1.70e+03
=====

```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.7e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```

[67]: #removed PF b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',

```

```

        'ORB', 'STL', 'BLK',
        'TS', 'DRBP', 'X3PAR',
        'ASTP', 'STLP', 'BLKP', 'OWS', 'DWS', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          SMOY      R-squared:                0.136
Model:                  OLS      Adj. R-squared:           0.129
Method:                 Least Squares      F-statistic:        18.47
Date:                  Wed, 12 May 2021    Prob (F-statistic):    3.09e-49
Time:                  19:19:03           Log-Likelihood:      2670.4
No. Observations:      1892             AIC:                -5307.
Df Residuals:          1875             BIC:                -5212.
Df Model:               16
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0412	0.018	2.348	0.019	0.007	0.076
Age	0.0005	0.000	1.383	0.167	-0.000	0.001
G	-0.0002	0.000	-1.453	0.146	-0.000	5.36e-05
GS	-0.0010	0.000	-5.493	0.000	-0.001	-0.001
X3PM	0.0299	0.004	6.767	0.000	0.021	0.039
X2PP	0.0644	0.027	2.392	0.017	0.012	0.117
ORB	-0.0011	0.004	-0.268	0.788	-0.009	0.007
STL	0.0144	0.008	1.750	0.080	-0.002	0.030
BLK	0.0121	0.009	1.361	0.174	-0.005	0.030
TS	-0.1336	0.029	-4.567	0.000	-0.191	-0.076
DRBP	-0.0006	0.000	-1.625	0.104	-0.001	0.000
X3PAR	-0.0474	0.010	-4.744	0.000	-0.067	-0.028
ASTP	-0.0001	0.000	-0.706	0.480	-0.001	0.000
STLP	-0.0046	0.003	-1.773	0.076	-0.010	0.000
BLKP	-0.0024	0.002	-1.355	0.176	-0.006	0.001
OWS	0.0115	0.002	7.015	0.000	0.008	0.015
DWS	0.0082	0.003	2.994	0.003	0.003	0.014

```

=====
Omnibus:                2899.560      Durbin-Watson:           1.749
Prob(Omnibus):           0.000      Jarque-Bera (JB):        994272.930
Skew:                    9.496      Prob(JB):                0.00

```


Kurtosis: 113.687 Cond. No. 1.69e+03

Warnings:

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

[2] The condition number is large, 1.69e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[68]: *#removed ASTP b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',  
         'X3PM', 'X2PP',  
         'ORB', 'STL', 'BLK',  
         'TS', 'DRBP', 'X3PAR',  
         'STLP', 'BLKP', 'OWS', 'DWS', ]  
x_train = train[cols] #all of the variables  
y_train = train['SMOY']  
#adding an intercept  
x_train = sm.add_constant(x_train)  
  
#fit data to the model  
model1= sm.OLS(y_train, x_train).fit()  
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY    R-squared:                0.136
Model:                  OLS     Adj. R-squared:           0.129
Method:                 Least Squares    F-statistic:        19.67
Date:                   Wed, 12 May 2021    Prob (F-statistic):    8.66e-50
Time:                   19:19:03    Log-Likelihood:        2670.1
No. Observations:      1892    AIC:                   -5308.
Df Residuals:          1876    BIC:                   -5220.
Df Model:               15
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0383	0.017	2.246	0.025	0.005	0.072
Age	0.0004	0.000	1.352	0.176	-0.000	0.001
G	-0.0002	0.000	-1.465	0.143	-0.000	5.24e-05
GS	-0.0010	0.000	-5.502	0.000	-0.001	-0.001
X3PM	0.0301	0.004	6.821	0.000	0.021	0.039
X2PP	0.0649	0.027	2.414	0.016	0.012	0.118
ORB	-0.0006	0.004	-0.142	0.887	-0.008	0.007
STL	0.0134	0.008	1.660	0.097	-0.002	0.029
BLK	0.0118	0.009	1.325	0.185	-0.006	0.029
TS	-0.1326	0.029	-4.539	0.000	-0.190	-0.075

DRBP	-0.0005	0.000	-1.522	0.128	-0.001	0.000
X3PAR	-0.0467	0.010	-4.700	0.000	-0.066	-0.027
STLP	-0.0047	0.003	-1.827	0.068	-0.010	0.000
BLKP	-0.0022	0.002	-1.252	0.211	-0.006	0.001
OWS	0.0114	0.002	6.980	0.000	0.008	0.015
DWS	0.0081	0.003	2.978	0.003	0.003	0.014

```
=====
Omnibus:                2900.142    Durbin-Watson:                1.750
Prob(Omnibus):           0.000    Jarque-Bera (JB):            994814.444
Skew:                    9.500    Prob(JB):                     0.00
Kurtosis:               113.717    Cond. No.                     1.67e+03
=====
```

Warnings:

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

[2] The condition number is large, 1.67e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[69]: *#removed BLK b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL',
         'TS', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'OWS', 'DWS', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          SMOY    R-squared:                0.135
Model:                  OLS     Adj. R-squared:           0.129
Method:                 Least Squares    F-statistic:            20.94
Date:                   Wed, 12 May 2021    Prob (F-statistic):      4.28e-50
Time:                   19:19:03    Log-Likelihood:         2669.2
No. Observations:       1892    AIC:                    -5308.
Df Residuals:           1877    BIC:                    -5225.
Df Model:                14
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]

const	0.0369	0.017	2.166	0.030	0.003	0.070
Age	0.0004	0.000	1.332	0.183	-0.000	0.001
G	-0.0002	0.000	-1.429	0.153	-0.000	5.62e-05
GS	-0.0009	0.000	-5.383	0.000	-0.001	-0.001
X3PM	0.0298	0.004	6.760	0.000	0.021	0.038
X2PP	0.0673	0.027	2.509	0.012	0.015	0.120
ORB	0.0020	0.003	0.590	0.555	-0.005	0.009
STL	0.0146	0.008	1.818	0.069	-0.001	0.030
TS	-0.1340	0.029	-4.588	0.000	-0.191	-0.077
DRBP	-0.0006	0.000	-1.793	0.073	-0.001	5.79e-05
X3PAR	-0.0443	0.010	-4.533	0.000	-0.064	-0.025
STLP	-0.0049	0.003	-1.900	0.058	-0.010	0.000
BLKP	-0.0005	0.001	-0.418	0.676	-0.003	0.002
OWS	0.0113	0.002	6.912	0.000	0.008	0.014
DWS	0.0086	0.003	3.153	0.002	0.003	0.014

Omnibus:	2902.994	Durbin-Watson:	1.749
Prob(Omnibus):	0.000	Jarque-Bera (JB):	999216.918
Skew:	9.518	Prob(JB):	0.00
Kurtosis:	113.963	Cond. No.	1.66e+03

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.66e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[70]: *#removed BLKP b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'STL',
         'TS', 'DRBP', 'X3PAR',
         'STLP', 'OWS', 'DWS', ]
x_train = train[cols] #all of the variables
y_train = train['SMOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	SMOY	R-squared:	0.135
Model:	OLS	Adj. R-squared:	0.129
Method:	Least Squares	F-statistic:	22.55

Date: Wed, 12 May 2021 Prob (F-statistic): 9.59e-51
Time: 19:19:03 Log-Likelihood: 2669.1
No. Observations: 1892 AIC: -5310.
Df Residuals: 1878 BIC: -5233.
Df Model: 13
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0375	0.017	2.211	0.027	0.004	0.071
Age	0.0004	0.000	1.328	0.184	-0.000	0.001
G	-0.0001	0.000	-1.393	0.164	-0.000	5.96e-05
GS	-0.0009	0.000	-5.380	0.000	-0.001	-0.001
X3PM	0.0297	0.004	6.754	0.000	0.021	0.038
X2PP	0.0662	0.027	2.481	0.013	0.014	0.119
ORB	0.0019	0.003	0.548	0.584	-0.005	0.009
STL	0.0149	0.008	1.861	0.063	-0.001	0.031
TS	-0.1354	0.029	-4.671	0.000	-0.192	-0.079
DRBP	-0.0007	0.000	-1.978	0.048	-0.001	-5.54e-06
X3PAR	-0.0434	0.010	-4.561	0.000	-0.062	-0.025
STLP	-0.0049	0.003	-1.899	0.058	-0.010	0.000
OWS	0.0114	0.002	7.053	0.000	0.008	0.015
DWS	0.0084	0.003	3.127	0.002	0.003	0.014

Omnibus: 2902.809 Durbin-Watson: 1.749
Prob(Omnibus): 0.000 Jarque-Bera (JB): 998835.238
Skew: 9.517 Prob(JB): 0.00
Kurtosis: 113.941 Cond. No. 1.66e+03

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.66e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[71]: import statsmodels.formula.api as smf

SMOYLinear = smf.ols(formula='SMOY ~ Age + G+ GS+
↳X3PM+X2PP+ORB+STL+TS+DRBP+X3PAR+STLP+OWS+DWS',
                    data=train).fit()
print(SMOYLinear.summary())
```

OLS Regression Results

Dep. Variable: SMOY R-squared: 0.135
Model: OLS Adj. R-squared: 0.129
Method: Least Squares F-statistic: 22.55

Date: Wed, 12 May 2021 Prob (F-statistic): 9.59e-51
Time: 19:19:03 Log-Likelihood: 2669.1
No. Observations: 1892 AIC: -5310.
Df Residuals: 1878 BIC: -5233.
Df Model: 13
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.0375	0.017	2.211	0.027	0.004	0.071
Age	0.0004	0.000	1.328	0.184	-0.000	0.001
G	-0.0001	0.000	-1.393	0.164	-0.000	5.96e-05
GS	-0.0009	0.000	-5.380	0.000	-0.001	-0.001
X3PM	0.0297	0.004	6.754	0.000	0.021	0.038
X2PP	0.0662	0.027	2.481	0.013	0.014	0.119
ORB	0.0019	0.003	0.548	0.584	-0.005	0.009
STL	0.0149	0.008	1.861	0.063	-0.001	0.031
TS	-0.1354	0.029	-4.671	0.000	-0.192	-0.079
DRBP	-0.0007	0.000	-1.978	0.048	-0.001	-5.54e-06
X3PAR	-0.0434	0.010	-4.561	0.000	-0.062	-0.025
STLP	-0.0049	0.003	-1.899	0.058	-0.010	0.000
OWS	0.0114	0.002	7.053	0.000	0.008	0.015
DWS	0.0084	0.003	3.127	0.002	0.003	0.014
Omnibus:	2902.809	Durbin-Watson:	1.749			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	998835.238			
Skew:	9.517	Prob(JB):	0.00			
Kurtosis:	113.941	Cond. No.	1.66e+03			

Warnings:

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

[2] The condition number is large, 1.66e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[72]: OSR2(SMOYLinear, train, test, 'SMOY')
```

```
[72]: 0.10762095053200527
```

Defensive Player of the Year

```
[163]: #Loading in dataset
dataset = pd.read_csv('update142.csv')
train = dataset[dataset['Season'] <= 2017]
test = dataset[dataset['Season'] > 2017]

y_train = train['MVP']
```

```

y_test = test['MVP']
x_train = train.iloc[:,6:51]
x_test = test.iloc[:,6:51]

```

```

[164]: #removed season b/c not relevant to prediciting our output variable
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

=====						
Dep. Variable:	DPOY	R-squared:	0.133			
Model:	OLS	Adj. R-squared:	0.126			
Method:	Least Squares	F-statistic:	21.33			
Date:	Wed, 12 May 2021	Prob (F-statistic):	1.90e-82			
Time:	19:31:06	Log-Likelihood:	5579.1			
No. Observations:	3231	AIC:	-1.111e+04			
Df Residuals:	3207	BIC:	-1.096e+04			
Df Model:	23					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.0106	0.013	0.811	0.418	-0.015	0.036
Age	-0.0003	0.000	-1.341	0.180	-0.001	0.000
G	-0.0002	5.9e-05	-3.218	0.001	-0.000	-7.41e-05
GS	-6.374e-05	4.32e-05	-1.475	0.140	-0.000	2.1e-05
X3PM	-0.0015	0.002	-0.701	0.483	-0.006	0.003
X3PP	0.0043	0.007	0.649	0.517	-0.009	0.017
X2PP	0.0142	0.017	0.846	0.398	-0.019	0.047
FTP	0.0040	0.007	0.559	0.576	-0.010	0.018
ORB	5.672e-05	0.002	0.031	0.976	-0.004	0.004
STL	0.0126	0.003	3.617	0.000	0.006	0.019
BLK	0.0296	0.003	8.676	0.000	0.023	0.036
PF	-0.0081	0.002	-4.587	0.000	-0.012	-0.005
TS	-0.0286	0.018	-1.560	0.119	-0.065	0.007

FTR	0.0290	0.007	4.373	0.000	0.016	0.042
DRBP	-0.0003	0.000	-1.570	0.117	-0.001	7.77e-05
X3PAR	0.0177	0.006	2.792	0.005	0.005	0.030
ASTP	-3.608e-05	0.000	-0.272	0.786	-0.000	0.000
STLP	-0.0044	0.002	-2.397	0.017	-0.008	-0.001
BLKP	-0.0019	0.001	-2.075	0.038	-0.004	-0.000
TOVP	0.0004	0.000	1.885	0.060	-1.54e-05	0.001
USG	-0.0002	0.000	-0.960	0.337	-0.001	0.000
OWS	0.0001	0.001	0.206	0.837	-0.001	0.001
DWS	0.0099	0.001	7.850	0.000	0.007	0.012
DBPM	0.0010	0.001	0.865	0.387	-0.001	0.003

```
=====
Omnibus:                    5939.334    Durbin-Watson:                1.107
Prob(Omnibus):              0.000    Jarque-Bera (JB):             7206173.730
Skew:                      13.579    Prob(JB):                     0.00
Kurtosis:                  232.761    Cond. No.                     2.32e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.32e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[202]: #removed ORB b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            DPOY    R-squared:                0.133
Model:                    OLS     Adj. R-squared:           0.127
Method:                    Least Squares    F-statistic:            22.31
Date:                    Wed, 12 May 2021    Prob (F-statistic):      3.98e-83
Time:                    19:40:21    Log-Likelihood:          5579.1
No. Observations:        3231    AIC:                    -1.111e+04
=====
```

Df Residuals: 3208 BIC: -1.097e+04
Df Model: 22
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0106	0.013	0.811	0.418	-0.015	0.036
Age	-0.0003	0.000	-1.343	0.179	-0.001	0.000
G	-0.0002	5.89e-05	-3.219	0.001	-0.000	-7.41e-05
GS	-6.355e-05	4.28e-05	-1.486	0.137	-0.000	2.03e-05
X3PM	-0.0015	0.002	-0.741	0.459	-0.005	0.002
X3PP	0.0042	0.006	0.651	0.515	-0.008	0.017
X2PP	0.0142	0.017	0.849	0.396	-0.019	0.047
FTP	0.0040	0.007	0.559	0.577	-0.010	0.018
STL	0.0126	0.003	3.620	0.000	0.006	0.019
BLK	0.0296	0.003	9.378	0.000	0.023	0.036
PF	-0.0081	0.002	-4.801	0.000	-0.011	-0.005
TS	-0.0286	0.018	-1.560	0.119	-0.065	0.007
FTR	0.0290	0.007	4.373	0.000	0.016	0.042
DRBP	-0.0003	0.000	-1.663	0.096	-0.001	5.54e-05
X3PAR	0.0177	0.006	2.792	0.005	0.005	0.030
ASTP	-3.639e-05	0.000	-0.275	0.783	-0.000	0.000
STLP	-0.0044	0.002	-2.404	0.016	-0.008	-0.001
BLKP	-0.0019	0.001	-2.105	0.035	-0.004	-0.000
TOVP	0.0004	0.000	1.888	0.059	-1.46e-05	0.001
USG	-0.0002	0.000	-0.964	0.335	-0.001	0.000
OWS	0.0001	0.001	0.215	0.830	-0.001	0.001
DWS	0.0099	0.001	7.854	0.000	0.007	0.012
DBPM	0.0010	0.001	0.876	0.381	-0.001	0.003
Omnibus:	5939.342	Durbin-Watson:	1.107			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	7206219.366			
Skew:	13.579	Prob(JB):	0.00			
Kurtosis:	232.762	Cond. No.	2.32e+03			

Warnings:

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

[2] The condition number is large, 2.32e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[203]: *#removed ASTP b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
```



```

        'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          DPOY      R-squared:                0.133
Model:                  OLS      Adj. R-squared:           0.127
Method:                 Least Squares      F-statistic:         23.37
Date:                   Wed, 12 May 2021    Prob (F-statistic):      8.42e-84
Time:                   19:40:39    Log-Likelihood:         5579.1
No. Observations:      3231    AIC:                    -1.111e+04
Df Residuals:          3209    BIC:                    -1.098e+04
Df Model:               21
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0105	0.013	0.803	0.422	-0.015	0.036
Age	-0.0003	0.000	-1.366	0.172	-0.001	0.000
G	-0.0002	5.89e-05	-3.233	0.001	-0.000	-7.49e-05
GS	-6.354e-05	4.28e-05	-1.486	0.137	-0.000	2.03e-05
X3PM	-0.0014	0.002	-0.713	0.476	-0.005	0.002
X3PP	0.0042	0.006	0.641	0.521	-0.009	0.017
X2PP	0.0144	0.017	0.861	0.389	-0.018	0.047
FTP	0.0040	0.007	0.556	0.578	-0.010	0.018
STL	0.0124	0.003	3.656	0.000	0.006	0.019
BLK	0.0297	0.003	9.395	0.000	0.023	0.036
PF	-0.0081	0.002	-4.794	0.000	-0.011	-0.005
TS	-0.0281	0.018	-1.539	0.124	-0.064	0.008
FTR	0.0292	0.007	4.461	0.000	0.016	0.042
DRBP	-0.0003	0.000	-1.661	0.097	-0.001	5.3e-05
X3PAR	0.0177	0.006	2.788	0.005	0.005	0.030
STLP	-0.0043	0.002	-2.390	0.017	-0.008	-0.001
BLKP	-0.0019	0.001	-2.087	0.037	-0.004	-0.000
TOVP	0.0004	0.000	2.064	0.039	1.76e-05	0.001
USG	-0.0003	0.000	-1.260	0.208	-0.001	0.000
OWS	8.45e-05	0.001	0.140	0.889	-0.001	0.001
DWS	0.0100	0.001	7.969	0.000	0.008	0.012
DBPM	0.0009	0.001	0.833	0.405	-0.001	0.003

```
=====
Omnibus:                    5939.300    Durbin-Watson:                    1.106
Prob(Omnibus):              0.000    Jarque-Bera (JB):              7205295.929
Skew:                      13.579    Prob(JB):                      0.00
Kurtosis:                  232.747    Cond. No.                      2.29e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.29e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[204]: *#removed DWS b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         'DBPM', ]
```

```
x_train = train[cols] #all of the variables
```

```
y_train = train['DPOY']
```

```
#adding an intercept
```

```
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
```

```
model1= sm.OLS(y_train, x_train).fit()
```

```
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            DPOY    R-squared:                0.133
Model:                    OLS     Adj. R-squared:          0.127
Method:                   Least Squares    F-statistic:            24.55
Date:                     Wed, 12 May 2021    Prob (F-statistic):      1.69e-84
Time:                     19:40:51    Log-Likelihood:          5579.1
No. Observations:         3231    AIC:                    -1.112e+04
Df Residuals:             3210    BIC:                    -1.099e+04
Df Model:                 20
Covariance Type:          nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0099	0.012	0.800	0.424	-0.014	0.034
Age	-0.0003	0.000	-1.360	0.174	-0.001	0.000
G	-0.0002	5.89e-05	-3.235	0.001	-0.000	-7.5e-05
GS	-6.252e-05	4.21e-05	-1.484	0.138	-0.000	2.01e-05
X3PM	-0.0014	0.002	-0.700	0.484	-0.005	0.002

X3PP	0.0041	0.006	0.638	0.524	-0.009	0.017
X2PP	0.0146	0.017	0.881	0.379	-0.018	0.047
FTP	0.0040	0.007	0.559	0.576	-0.010	0.018
STL	0.0124	0.003	3.664	0.000	0.006	0.019
BLK	0.0297	0.003	9.402	0.000	0.023	0.036
PF	-0.0081	0.002	-4.843	0.000	-0.011	-0.005
TS	-0.0273	0.017	-1.571	0.116	-0.061	0.007
FTR	0.0294	0.006	4.565	0.000	0.017	0.042
DRBP	-0.0003	0.000	-1.666	0.096	-0.001	5.21e-05
X3PAR	0.0176	0.006	2.785	0.005	0.005	0.030
STLP	-0.0043	0.002	-2.397	0.017	-0.008	-0.001
BLKP	-0.0019	0.001	-2.109	0.035	-0.004	-0.000
TOVP	0.0003	0.000	2.065	0.039	1.76e-05	0.001
USG	-0.0003	0.000	-1.256	0.209	-0.001	0.000
DWS	0.0100	0.001	8.410	0.000	0.008	0.012
DBPM	0.0009	0.001	0.829	0.407	-0.001	0.003

```
=====
Omnibus:                    5939.068    Durbin-Watson:                1.107
Prob(Omnibus):              0.000    Jarque-Bera (JB):              7202919.454
Skew:                      13.578    Prob(JB):                      0.00
Kurtosis:                  232.709    Cond. No.                      2.26e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.26e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[205]: #removed FTP b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:              DPOY    R-squared:                0.133
```

```

Model:                                OLS      Adj. R-squared:            0.127
Method:                             Least Squares      F-statistic:            25.83
Date:                               Wed, 12 May 2021    Prob (F-statistic):      3.81e-85
Time:                               19:41:15          Log-Likelihood:          5578.9
No. Observations:                   3231          AIC:                    -1.112e+04
Df Residuals:                       3211          BIC:                    -1.100e+04
Df Model:                           19
Covariance Type:                    nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0118	0.012	0.998	0.318	-0.011	0.035
Age	-0.0002	0.000	-1.334	0.182	-0.001	0.000
G	-0.0002	5.85e-05	-3.195	0.001	-0.000	-7.23e-05
GS	-6.278e-05	4.21e-05	-1.490	0.136	-0.000	1.98e-05
X3PM	-0.0013	0.002	-0.646	0.518	-0.005	0.003
X3PP	0.0045	0.006	0.701	0.483	-0.008	0.017
X2PP	0.0132	0.016	0.807	0.420	-0.019	0.045
STL	0.0124	0.003	3.679	0.000	0.006	0.019
BLK	0.0297	0.003	9.394	0.000	0.023	0.036
PF	-0.0081	0.002	-4.838	0.000	-0.011	-0.005
TS	-0.0251	0.017	-1.484	0.138	-0.058	0.008
FTR	0.0292	0.006	4.538	0.000	0.017	0.042
DRBP	-0.0003	0.000	-1.717	0.086	-0.001	4.3e-05
X3PAR	0.0175	0.006	2.766	0.006	0.005	0.030
STLP	-0.0044	0.002	-2.443	0.015	-0.008	-0.001
BLKP	-0.0019	0.001	-2.151	0.032	-0.004	-0.000
TOVP	0.0003	0.000	2.050	0.040	1.51e-05	0.001
USG	-0.0002	0.000	-1.164	0.244	-0.001	0.000
DWS	0.0100	0.001	8.394	0.000	0.008	0.012
DBPM	0.0009	0.001	0.853	0.394	-0.001	0.003

```

Omnibus:                            5939.496    Durbin-Watson:            1.106
Prob(Omnibus):                      0.000      Jarque-Bera (JB):        7206095.273
Skew:                               13.580      Prob(JB):                0.00
Kurtosis:                           232.760      Cond. No.                2.21e+03

```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.21e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```

[206]: #removed X3PM b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'X3PP', 'X2PP',

```

```

        'STL', 'BLK',
        'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
        'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
        'DBPM', ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

Dep. Variable:	DPOY	R-squared:	0.132
Model:	OLS	Adj. R-squared:	0.128
Method:	Least Squares	F-statistic:	27.24
Date:	Wed, 12 May 2021	Prob (F-statistic):	8.78e-86
Time:	19:41:27	Log-Likelihood:	5578.7
No. Observations:	3231	AIC:	-1.112e+04
Df Residuals:	3212	BIC:	-1.100e+04
Df Model:	18		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0136	0.012	1.176	0.240	-0.009	0.036
Age	-0.0003	0.000	-1.379	0.168	-0.001	0.000
G	-0.0002	5.83e-05	-3.263	0.001	-0.000	-7.6e-05
GS	-6.744e-05	4.15e-05	-1.626	0.104	-0.000	1.39e-05
X3PP	0.0035	0.006	0.554	0.579	-0.009	0.016
X2PP	0.0143	0.016	0.877	0.381	-0.018	0.046
STL	0.0118	0.003	3.637	0.000	0.005	0.018
BLK	0.0302	0.003	9.921	0.000	0.024	0.036
PF	-0.0082	0.002	-4.914	0.000	-0.011	-0.005
TS	-0.0268	0.017	-1.604	0.109	-0.060	0.006
FTR	0.0290	0.006	4.513	0.000	0.016	0.042
DRBP	-0.0003	0.000	-1.710	0.087	-0.001	4.41e-05
X3PAR	0.0153	0.005	2.857	0.004	0.005	0.026
STLP	-0.0043	0.002	-2.383	0.017	-0.008	-0.001
BLKP	-0.0020	0.001	-2.342	0.019	-0.004	-0.000
TOVP	0.0003	0.000	2.059	0.040	1.65e-05	0.001
USG	-0.0003	0.000	-1.344	0.179	-0.001	0.000
DWS	0.0100	0.001	8.412	0.000	0.008	0.012
DBPM	0.0009	0.001	0.894	0.371	-0.001	0.003

Omnibus:	5939.149	Durbin-Watson:	1.106
Prob(Omnibus):	0.000	Jarque-Bera (JB):	7202996.725
Skew:	13.578	Prob(JB):	0.00
Kurtosis:	232.710	Cond. No.	2.19e+03

=====

Warnings:

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

[2] The condition number is large, 2.19e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[207]: *#removed X3PP b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',
         'X2PP',
         'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         'DBPM', ]
```

```
x_train = train[cols] #all of the variables
```

```
y_train = train['DPOY']
```

```
#adding an intercept
```

```
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
```

```
model1= sm.OLS(y_train, x_train).fit()
```

```
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          DPOY      R-squared:                0.132
Model:                  OLS      Adj. R-squared:           0.128
Method:                 Least Squares      F-statistic:          28.83
Date:                   Wed, 12 May 2021    Prob (F-statistic):      1.87e-86
Time:                   19:41:37            Log-Likelihood:        5578.6
No. Observations:      3231              AIC:                  -1.112e+04
Df Residuals:          3213              BIC:                  -1.101e+04
Df Model:               17
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0139	0.012	1.207	0.228	-0.009	0.037
Age	-0.0003	0.000	-1.378	0.168	-0.001	0.000
G	-0.0002	5.8e-05	-3.225	0.001	-0.000	-7.34e-05
GS	-6.623e-05	4.14e-05	-1.599	0.110	-0.000	1.5e-05
X2PP	0.0129	0.016	0.802	0.422	-0.019	0.045
STL	0.0121	0.003	3.746	0.000	0.006	0.018

BLK	0.0300	0.003	9.916	0.000	0.024	0.036
PF	-0.0083	0.002	-4.989	0.000	-0.012	-0.005
TS	-0.0253	0.016	-1.534	0.125	-0.058	0.007
FTR	0.0286	0.006	4.480	0.000	0.016	0.041
DRBP	-0.0003	0.000	-1.736	0.083	-0.001	3.95e-05
X3PAR	0.0163	0.005	3.196	0.001	0.006	0.026
STLP	-0.0043	0.002	-2.398	0.017	-0.008	-0.001
BLKP	-0.0020	0.001	-2.334	0.020	-0.004	-0.000
TOVP	0.0003	0.000	2.051	0.040	1.53e-05	0.001
USG	-0.0002	0.000	-1.285	0.199	-0.001	0.000
DWS	0.0100	0.001	8.402	0.000	0.008	0.012
DBPM	0.0009	0.001	0.900	0.368	-0.001	0.003

```
=====
Omnibus:                    5938.747    Durbin-Watson:                1.106
Prob(Omnibus):              0.000    Jarque-Bera (JB):             7197341.968
Skew:                      13.576    Prob(JB):                     0.00
Kurtosis:                  232.619    Cond. No.                     2.15e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.15e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[208]: *#removed X2PP b/c not statistically significant*

```
cols = [ 'Age', 'G', 'GS',

        'STL', 'BLK',
        'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
        'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
        'DBPM', ]

x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          DPOY    R-squared:                0.132
Model:                  OLS    Adj. R-squared:         0.128
Method:                 Least Squares    F-statistic:          30.60
Date:                   Wed, 12 May 2021    Prob (F-statistic):    4.53e-87
Time:                   19:41:50    Log-Likelihood:        5578.2
=====
```

No. Observations: 3231 AIC: -1.112e+04
Df Residuals: 3214 BIC: -1.102e+04
Df Model: 16
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0166	0.011	1.502	0.133	-0.005	0.038
Age	-0.0003	0.000	-1.383	0.167	-0.001	0.000
G	-0.0002	5.78e-05	-3.172	0.002	-0.000	-7.01e-05
GS	-6.634e-05	4.14e-05	-1.601	0.109	-0.000	1.49e-05
STL	0.0121	0.003	3.751	0.000	0.006	0.018
BLK	0.0305	0.003	10.280	0.000	0.025	0.036
PF	-0.0083	0.002	-5.013	0.000	-0.012	-0.005
TS	-0.0185	0.014	-1.307	0.191	-0.046	0.009
FTR	0.0284	0.006	4.450	0.000	0.016	0.041
DRBP	-0.0003	0.000	-1.676	0.094	-0.001	4.99e-05
X3PAR	0.0156	0.005	3.106	0.002	0.006	0.025
STLP	-0.0042	0.002	-2.366	0.018	-0.008	-0.001
BLKP	-0.0021	0.001	-2.360	0.018	-0.004	-0.000
TOVP	0.0003	0.000	2.044	0.041	1.41e-05	0.001
USG	-0.0003	0.000	-1.391	0.164	-0.001	0.000
DWS	0.0100	0.001	8.378	0.000	0.008	0.012
DBPM	0.0009	0.001	0.856	0.392	-0.001	0.003
Omnibus:	5938.807		Durbin-Watson:	1.107		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	7195953.930		
Skew:	13.577		Prob(JB):	0.00		
Kurtosis:	232.596		Cond. No.	1.73e+03		

Warnings:

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

[2] The condition number is large, 1.73e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[209]: #removed DBPM b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)
```



```
#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          DPOY      R-squared:                0.132
Model:                  OLS       Adj. R-squared:           0.128
Method:                 Least Squares   F-statistic:           32.59
Date:                  Wed, 12 May 2021   Prob (F-statistic):    1.11e-87
Time:                  19:42:03    Log-Likelihood:        5577.9
No. Observations:      3231      AIC:                   -1.112e+04
Df Residuals:          3215      BIC:                   -1.103e+04
Df Model:               15
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0135	0.010	1.295	0.195	-0.007	0.034
Age	-0.0002	0.000	-1.296	0.195	-0.001	0.000
G	-0.0002	5.73e-05	-3.316	0.001	-0.000	-7.77e-05
GS	-6.745e-05	4.14e-05	-1.629	0.103	-0.000	1.37e-05
STL	0.0120	0.003	3.715	0.000	0.006	0.018
BLK	0.0303	0.003	10.245	0.000	0.025	0.036
PF	-0.0083	0.002	-5.039	0.000	-0.012	-0.005
TS	-0.0148	0.013	-1.100	0.271	-0.041	0.012
FTR	0.0276	0.006	4.373	0.000	0.015	0.040
DRBP	-0.0003	0.000	-1.609	0.108	-0.001	6.14e-05
X3PAR	0.0159	0.005	3.177	0.002	0.006	0.026
STLP	-0.0033	0.001	-2.294	0.022	-0.006	-0.000
BLKP	-0.0017	0.001	-2.204	0.028	-0.003	-0.000
TOVP	0.0003	0.000	2.062	0.039	1.71e-05	0.001
USG	-0.0003	0.000	-1.983	0.047	-0.001	-3.87e-06
DWS	0.0104	0.001	9.823	0.000	0.008	0.012

```
=====
Omnibus:                5939.023    Durbin-Watson:           1.108
Prob(Omnibus):           0.000      Jarque-Bera (JB):        7194613.321
Skew:                    13.578      Prob(JB):                0.00
Kurtosis:                232.574     Cond. No.                 1.62e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.62e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[210]: #removed TS b/c not statistically significant
cols = [ 'Age', 'G', 'GS',
         'STL', 'BLK',
         'PF', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          DPOY      R-squared:                0.132
Model:                  OLS      Adj. R-squared:           0.128
Method:                 Least Squares      F-statistic:        34.83
Date:                   Wed, 12 May 2021    Prob (F-statistic):    3.32e-88
Time:                   19:42:15           Log-Likelihood:       5577.3
No. Observations:      3231           AIC:                  -1.112e+04
Df Residuals:          3216           BIC:                  -1.103e+04
Df Model:               14
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0072	0.009	0.827	0.408	-0.010	0.024
Age	-0.0002	0.000	-1.321	0.187	-0.001	0.000
G	-0.0002	5.7e-05	-3.444	0.001	-0.000	-8.46e-05
GS	-6.707e-05	4.14e-05	-1.620	0.105	-0.000	1.41e-05
STL	0.0120	0.003	3.717	0.000	0.006	0.018
BLK	0.0305	0.003	10.341	0.000	0.025	0.036
PF	-0.0085	0.002	-5.181	0.000	-0.012	-0.005
FTR	0.0255	0.006	4.239	0.000	0.014	0.037
DRBP	-0.0003	0.000	-1.600	0.110	-0.001	6.31e-05
X3PAR	0.0149	0.005	3.024	0.003	0.005	0.025
STLP	-0.0032	0.001	-2.218	0.027	-0.006	-0.000
BLKP	-0.0018	0.001	-2.338	0.019	-0.003	-0.000
TOVP	0.0004	0.000	2.206	0.027	4.1e-05	0.001
USG	-0.0003	0.000	-1.975	0.048	-0.001	-2.4e-06
DWS	0.0102	0.001	9.770	0.000	0.008	0.012

```
=====
Omnibus:                 5941.621    Durbin-Watson:           1.108
Prob(Omnibus):           0.000      Jarque-Bera (JB):        7218470.187
=====
```

Skew:	13.590	Prob(JB):	0.00
Kurtosis:	232.957	Cond. No.	979.

=====

Warnings:

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

```
[211]: #removed Age b/c not statistically significant
cols = [ 'G', 'GS',
         'STL', 'BLK',
         'PF', 'FTR', 'DRBP', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          DPOY      R-squared:                0.131
Model:                  OLS       Adj. R-squared:         0.128
Method:                 Least Squares   F-statistic:          37.37
Date:                   Wed, 12 May 2021   Prob (F-statistic):    1.25e-88
Time:                   19:42:29    Log-Likelihood:        5576.4
No. Observations:      3231    AIC:                   -1.112e+04
Df Residuals:          3217    BIC:                   -1.104e+04
Df Model:               13
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0003	0.007	0.036	0.971	-0.013	0.014
G	-0.0002	5.69e-05	-3.354	0.001	-0.000	-7.92e-05
GS	-6.983e-05	4.14e-05	-1.689	0.091	-0.000	1.13e-05
STL	0.0122	0.003	3.784	0.000	0.006	0.018
BLK	0.0308	0.003	10.455	0.000	0.025	0.037
PF	-0.0086	0.002	-5.200	0.000	-0.012	-0.005
FTR	0.0259	0.006	4.315	0.000	0.014	0.038
DRBP	-0.0003	0.000	-1.633	0.103	-0.001	5.72e-05
X3PAR	0.0144	0.005	2.933	0.003	0.005	0.024
STLP	-0.0032	0.001	-2.194	0.028	-0.006	-0.000
BLKP	-0.0018	0.001	-2.341	0.019	-0.003	-0.000

TOVP	0.0004	0.000	2.160	0.031	3.34e-05	0.001
USG	-0.0003	0.000	-1.848	0.065	-0.001	1.92e-05
DWS	0.0101	0.001	9.680	0.000	0.008	0.012

```
=====
Omnibus:                    5944.767    Durbin-Watson:                1.107
Prob(Omnibus):              0.000    Jarque-Bera (JB):            7240721.920
Skew:                      13.605    Prob(JB):                   0.00
Kurtosis:                  233.313    Cond. No.                   809.
=====
```

Warnings:

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

```
[212]: #removed DRBP b/c not statistically significant
cols = [ 'G', 'GS',
         'STL', 'BLK',
         'PF', 'FTR', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            DPOY    R-squared:                0.130
Model:                    OLS     Adj. R-squared:         0.127
Method:                   Least Squares    F-statistic:          40.24
Date:                     Wed, 12 May 2021    Prob (F-statistic):    7.06e-89
Time:                     19:42:44    Log-Likelihood:        5575.0
No. Observations:        3231    AIC:                   -1.112e+04
Df Residuals:            3218    BIC:                   -1.105e+04
Df Model:                12
Covariance Type:         nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0042	0.006	-0.647	0.518	-0.017	0.008
G	-0.0002	5.63e-05	-3.150	0.002	-0.000	-6.7e-05
GS	-6.8e-05	4.14e-05	-1.645	0.100	-0.000	1.31e-05
STL	0.0128	0.003	3.985	0.000	0.006	0.019
BLK	0.0309	0.003	10.486	0.000	0.025	0.037

PF	-0.0090	0.002	-5.511	0.000	-0.012	-0.006
FTR	0.0257	0.006	4.286	0.000	0.014	0.038
X3PAR	0.0165	0.005	3.478	0.001	0.007	0.026
STLP	-0.0030	0.001	-2.055	0.040	-0.006	-0.000
BLKP	-0.0022	0.001	-2.838	0.005	-0.004	-0.001
TOVP	0.0004	0.000	2.123	0.034	2.72e-05	0.001
USG	-0.0003	0.000	-1.783	0.075	-0.001	3.02e-05
DWS	0.0095	0.001	9.659	0.000	0.008	0.011

```
=====
Omnibus:                    5944.119    Durbin-Watson:                1.106
Prob(Omnibus):              0.000    Jarque-Bera (JB):              7231847.314
Skew:                      13.602    Prob(JB):                      0.00
Kurtosis:                  233.170    Cond. No.                      754.
=====
```

Warnings:

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

```
[213]: #removed GS b/c not statistically significant
cols = [ 'G',
         'STL', 'BLK',
         'PF', 'FTR', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            DPOY    R-squared:                0.130
Model:                    OLS     Adj. R-squared:           0.127
Method:                    Least Squares    F-statistic:            43.63
Date:                      Wed, 12 May 2021    Prob (F-statistic):      3.92e-89
Time:                      19:43:01    Log-Likelihood:          5573.7
No. Observations:          3231    AIC:                     -1.112e+04
Df Residuals:              3219    BIC:                     -1.105e+04
Df Model:                  11
Covariance Type:           nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	-0.0033	0.006	-0.514	0.607	-0.016	0.009
G	-0.0002	5.47e-05	-3.644	0.000	-0.000	-9.21e-05
STL	0.0105	0.003	3.629	0.000	0.005	0.016
BLK	0.0299	0.003	10.366	0.000	0.024	0.036
PF	-0.0093	0.002	-5.754	0.000	-0.012	-0.006
FTR	0.0261	0.006	4.348	0.000	0.014	0.038
X3PAR	0.0169	0.005	3.579	0.000	0.008	0.026
STLP	-0.0023	0.001	-1.674	0.094	-0.005	0.000
BLKP	-0.0020	0.001	-2.617	0.009	-0.003	-0.000
TOVP	0.0004	0.000	2.133	0.033	2.88e-05	0.001
USG	-0.0003	0.000	-1.849	0.065	-0.001	1.9e-05
DWS	0.0093	0.001	9.519	0.000	0.007	0.011

```
=====
Omnibus:                    5947.654    Durbin-Watson:                1.104
Prob(Omnibus):              0.000    Jarque-Bera (JB):             7260835.205
Skew:                      13.619    Prob(JB):                     0.00
Kurtosis:                  233.634    Cond. No.                     670.
=====
```

Warnings:

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

```
[214]: #removed STLP b/c not statistically significant
cols = [ 'G',
         'STL', 'BLK',
         'PF', 'FTR', 'X3PAR',
         'BLKP', 'TOVP', 'USG', 'DWS',
         ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            DPOY    R-squared:                0.129
Model:                    OLS     Adj. R-squared:           0.126
Method:                    Least Squares    F-statistic:              47.68
Date:                      Wed, 12 May 2021    Prob (F-statistic):       2.18e-89
Time:                      19:43:15    Log-Likelihood:           5572.3
No. Observations:          3231    AIC:                     -1.112e+04
Df Residuals:              3220    BIC:                     -1.106e+04
Df Model:                  10
```

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0069	0.006	-1.141	0.254	-0.019	0.005
G	-0.0002	5.41e-05	-3.427	0.001	-0.000	-7.93e-05
STL	0.0079	0.002	3.235	0.001	0.003	0.013
BLK	0.0301	0.003	10.423	0.000	0.024	0.036
PF	-0.0088	0.002	-5.539	0.000	-0.012	-0.006
FTR	0.0259	0.006	4.314	0.000	0.014	0.038
X3PAR	0.0171	0.005	3.613	0.000	0.008	0.026
BLKP	-0.0019	0.001	-2.574	0.010	-0.003	-0.000
TOVP	0.0003	0.000	1.952	0.051	-1.43e-06	0.001
USG	-0.0003	0.000	-1.748	0.081	-0.001	3.62e-05
DWS	0.0092	0.001	9.462	0.000	0.007	0.011
=====						
Omnibus:		5948.044	Durbin-Watson:		1.105	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		7252383.826	
Skew:		13.622	Prob(JB):		0.00	
Kurtosis:		233.497	Cond. No.		654.	
=====						

Warnings:

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

```
[216]: #removed USG b/c not statistically significant
cols = [ 'G',
         'STL', 'BLK',
         'PF', 'FTR', 'X3PAR',
         'BLKP', 'TOVP', 'DWS',
       ]
x_train = train[cols] #all of the variables
y_train = train['DPOY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	DPOY	R-squared:	0.128
Model:	OLS	Adj. R-squared:	0.126
Method:	Least Squares	F-statistic:	52.61
Date:	Wed, 12 May 2021	Prob (F-statistic):	1.31e-89
Time:	19:43:39	Log-Likelihood:	5570.7

```

No. Observations:      3231    AIC:                -1.112e+04
Df Residuals:          3221    BIC:                -1.106e+04
Df Model:              9
Covariance Type:      nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         -0.0133      0.005      -2.753      0.006      -0.023      -0.004
G              -0.0002     5.4e-05     -3.330      0.001      -0.000     -7.39e-05
STL            0.0071      0.002       2.958      0.003       0.002       0.012
BLK            0.0298      0.003     10.341      0.000       0.024       0.035
PF            -0.0089      0.002     -5.570      0.000      -0.012      -0.006
FTR            0.0256      0.006       4.263      0.000       0.014       0.037
X3PAR          0.0182      0.005       3.882      0.000       0.009       0.027
BLKP          -0.0017      0.001     -2.281      0.023      -0.003      -0.000
TOVP           0.0004      0.000       2.386      0.017      6.9e-05       0.001
DWS            0.0091      0.001       9.331      0.000       0.007       0.011
=====

Omnibus:                5949.356    Durbin-Watson:                1.105
Prob(Omnibus):           0.000    Jarque-Bera (JB):            7255172.866
Skew:                    13.629    Prob(JB):                     0.00
Kurtosis:                233.540    Cond. No.                     598.
=====

```

Warnings:

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

```

[217]: import statsmodels.formula.api as smf

DPOYLinear = smf.ols(formula='DPOY ~ G + STL + BLK + PF + FTR + X3PAR+ BLKP +_
↪TOVP + DWS',
                      data=train).fit()
print(DPOYLinear.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          DPOY    R-squared:                0.128
Model:                  OLS    Adj. R-squared:          0.126
Method:                 Least Squares    F-statistic:            52.61
Date:                   Wed, 12 May 2021    Prob (F-statistic):      1.31e-89
Time:                   19:44:36    Log-Likelihood:          5570.7
No. Observations:       3231    AIC:                    -1.112e+04
Df Residuals:           3221    BIC:                    -1.106e+04
Df Model:               9
Covariance Type:        nonrobust
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----

```



```
-----
Intercept    -0.0133    0.005    -2.753    0.006    -0.023    -0.004
G            -0.0002    5.4e-05   -3.330    0.001    -0.000    -7.39e-05
STL          0.0071    0.002     2.958    0.003    0.002     0.012
BLK          0.0298    0.003    10.341    0.000    0.024     0.035
PF           -0.0089    0.002    -5.570    0.000    -0.012    -0.006
FTR          0.0256    0.006     4.263    0.000    0.014     0.037
X3PAR        0.0182    0.005     3.882    0.000    0.009     0.027
BLKP         -0.0017    0.001    -2.281    0.023    -0.003    -0.000
TOVP         0.0004    0.000     2.386    0.017    6.9e-05    0.001
DWS          0.0091    0.001     9.331    0.000    0.007     0.011
=====
Omnibus:                5949.356    Durbin-Watson:                1.105
Prob(Omnibus):           0.000    Jarque-Bera (JB):             7255172.866
Skew:                    13.629    Prob(JB):                     0.00
Kurtosis:                233.540    Cond. No.                     598.
=====
```

Warnings:

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

```
[218]: OSR2(DPOYLinear, train, test, 'DPOY')
```

```
[218]: 0.11858601031363625
```

MIP

```
[184]: #loading in the MIP data
datasetmip = pd.read_csv('mip_stats.csv')
trainmip = datasetmip[datasetmip['Season'] <= 2017]
testmip = datasetmip[datasetmip['Season'] > 2017]
testmip
```

```
[184]:
```

	Player	Season	G	GS	MP	FGM	FGA	FGP	X3PM	X3PA	\
11	Aaron Brooks	2018.0	-33.0	1.0	-7.9	-1.0	-2.4	0.003	-0.4	-1.0	
14	Aaron Gordon	2018.0	-22.0	-15.0	4.2	1.6	4.1	-0.020	1.0	2.6	
15	Aaron Gordon	2019.0	20.0	21.0	0.9	-0.5	-1.5	0.015	-0.4	-1.5	
16	Aaron Gordon	2020.0	-16.0	-16.0	-1.3	-0.6	-1.0	-0.012	-0.4	-0.6	
21	Aaron Holiday	2020.0	16.0	33.0	11.6	1.4	3.3	0.013	0.4	0.8	
...	
3390	Zach LaVine	2019.0	39.0	38.0	7.2	2.7	3.2	0.084	0.1	0.0	
3391	Zach LaVine	2020.0	-3.0	-2.0	0.3	0.6	2.0	-0.017	1.2	3.0	
3399	Zach Randolph	2018.0	-14.0	52.0	1.1	0.2	-0.3	0.024	0.6	1.2	
3407	Zaza Pachulia	2018.0	-1.0	-13.0	-4.0	-0.1	-0.6	0.030	0.0	0.0	
3408	Zaza Pachulia	2019.0	-1.0	-54.0	-1.2	-0.9	-1.0	-0.124	0.0	0.1	
...	
	USG	OVS	DWS	WS	WS48	OBPM	DBPM	BPM	VORP	MIP	

```

11    ...  0.7  0.3 -0.4 -0.2  0.017  -0.2  -0.1 -0.1   0.3  0.000
14    ...  4.6 -1.1  0.3 -0.8 -0.005   1.2   0.1  1.3   0.5  0.004
15    ... -2.9  0.9  1.3  2.2  0.021  -0.6   0.9  0.3   0.6  0.000
16    ... -1.1 -0.4 -1.0 -1.4 -0.006  -0.4  -0.2 -0.6  -0.7  0.000
21    ... -3.2  0.3  0.9  1.2 -0.002   0.1  -0.1  0.0   0.1  0.000

...    ...    ...    ...    ...    ...    ...    ...    ...    ...
3390  ...  1.0  1.8  0.6  2.5  0.037   3.2  -0.1  3.1   1.6  0.002
3391  ...  1.2  0.1  1.1  1.2  0.029   0.8   0.8  1.6   0.9  0.000
3399  ... -1.6 -0.5 -1.2 -1.7 -0.038  -0.5   0.1 -0.5  -0.3  0.000
3407  ...  1.7 -0.5 -1.1 -1.4 -0.017   0.4  -0.9 -0.6  -0.3  0.000
3408  ... -2.0 -0.7 -0.1 -1.0 -0.033  -2.1   0.9 -1.2  -0.3  0.000

```

[1020 rows x 48 columns]

```

[93]: #removed age b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          MIP      R-squared:                0.071
Model:                  OLS      Adj. R-squared:           0.062
Method:                 Least Squares      F-statistic:          8.225
Date:                   Wed, 12 May 2021    Prob (F-statistic):      7.84e-26
Time:                   19:19:05           Log-Likelihood:         4147.9
No. Observations:       2390             AIC:                  -8250.
Df Residuals:           2367             BIC:                  -8117.
Df Model:                22
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.036	0.000	0.004	0.007
G	-7.104e-05	5.54e-05	-1.282	0.200	-0.000	3.76e-05
GS	8.422e-05	4.49e-05	1.877	0.061	-3.76e-06	0.000

X3PM	0.0142	0.003	5.008	0.000	0.009	0.020
X3PP	-0.0025	0.006	-0.400	0.689	-0.015	0.010
X2PP	0.0146	0.017	0.869	0.385	-0.018	0.048
FTP	0.0052	0.008	0.633	0.527	-0.011	0.021
ORB	0.0141	0.003	5.310	0.000	0.009	0.019
STL	0.0018	0.005	0.387	0.699	-0.007	0.011
BLK	0.0160	0.004	3.587	0.000	0.007	0.025
PF	-0.0095	0.002	-3.824	0.000	-0.014	-0.005
TS	-0.0243	0.015	-1.599	0.110	-0.054	0.005
FTR	-1.122e-05	0.008	-0.001	0.999	-0.015	0.015
DRBP	-5.37e-05	0.000	-0.264	0.792	-0.000	0.000
X3PAR	-0.0114	0.006	-1.766	0.077	-0.024	0.001
ASTP	-7.935e-05	0.000	-0.545	0.586	-0.000	0.000
STLP	-0.0039	0.002	-2.289	0.022	-0.007	-0.001
BLKP	-0.0024	0.001	-2.932	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.130	0.033	3.2e-05	0.001
USG	0.0009	0.000	3.340	0.001	0.000	0.001
OWS	0.0024	0.001	3.729	0.000	0.001	0.004
DWS	-0.0020	0.001	-1.853	0.064	-0.004	0.000
DBPM	0.0020	0.001	1.886	0.059	-7.83e-05	0.004

```
=====
Omnibus:                4360.865    Durbin-Watson:                1.954
Prob(Omnibus):           0.000    Jarque-Bera (JB):            4780912.156
Skew:                    13.266    Prob(JB):                     0.00
Kurtosis:                220.498    Cond. No.                     587.
=====
```

Warnings:

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

```
[94]: #removed ftr not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```

=====
Dep. Variable:          MIP    R-squared:          0.071
Model:                 OLS    Adj. R-squared:       0.063
Method:                Least Squares    F-statistic:        8.620
Date:                  Wed, 12 May 2021    Prob (F-statistic):   2.62e-26
Time:                  19:19:06    Log-Likelihood:       4147.9
No. Observations:      2390    AIC:                  -8252.
Df Residuals:          2368    BIC:                  -8125.
Df Model:              21
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.041	0.000	0.004	0.007
G	-7.103e-05	5.53e-05	-1.284	0.199	-0.000	3.75e-05
GS	8.422e-05	4.48e-05	1.880	0.060	-3.62e-06	0.000
X3PM	0.0142	0.003	5.009	0.000	0.009	0.020
X3PP	-0.0025	0.006	-0.401	0.689	-0.015	0.010
X2PP	0.0146	0.017	0.872	0.383	-0.018	0.048
FTP	0.0052	0.008	0.633	0.527	-0.011	0.021
ORB	0.0141	0.003	5.320	0.000	0.009	0.019
STL	0.0018	0.005	0.387	0.698	-0.007	0.011
BLK	0.0160	0.004	3.597	0.000	0.007	0.025
PF	-0.0095	0.002	-3.826	0.000	-0.014	-0.005
TS	-0.0243	0.015	-1.629	0.103	-0.054	0.005
DRBP	-5.37e-05	0.000	-0.264	0.792	-0.000	0.000
X3PAR	-0.0114	0.006	-1.890	0.059	-0.023	0.000
ASTP	-7.933e-05	0.000	-0.546	0.585	-0.000	0.000
STLP	-0.0039	0.002	-2.309	0.021	-0.007	-0.001
BLKP	-0.0024	0.001	-2.965	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.138	0.033	3.34e-05	0.001
USG	0.0009	0.000	3.341	0.001	0.000	0.001
OWS	0.0024	0.001	3.771	0.000	0.001	0.004
DWS	-0.0020	0.001	-1.856	0.064	-0.004	0.000
DBPM	0.0020	0.001	1.896	0.058	-6.72e-05	0.004

```

=====
Omnibus:              4360.868    Durbin-Watson:          1.954
Prob(Omnibus):        0.000    Jarque-Bera (JB):       4780931.770
Skew:                 13.266    Prob(JB):               0.00
Kurtosis:             220.498    Cond. No.               578.
=====

```

Warnings:

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

```
[95]: #removed DRBP not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
```

Dep. Variable:	MIP	R-squared:	0.071
Model:	OLS	Adj. R-squared:	0.063
Method:	Least Squares	F-statistic:	9.051
Date:	Wed, 12 May 2021	Prob (F-statistic):	8.83e-27
Time:	19:19:06	Log-Likelihood:	4147.9
No. Observations:	2390	AIC:	-8254.
Df Residuals:	2369	BIC:	-8132.
Df Model:	20		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.0054	0.001	6.038	0.000	0.004	0.007
G	-7.006e-05	5.52e-05	-1.269	0.205	-0.000	3.82e-05
GS	8.427e-05	4.48e-05	1.882	0.060	-3.56e-06	0.000
X3PM	0.0142	0.003	5.005	0.000	0.009	0.020
X3PP	-0.0024	0.006	-0.393	0.694	-0.015	0.010
X2PP	0.0146	0.017	0.869	0.385	-0.018	0.047
FTP	0.0052	0.008	0.635	0.525	-0.011	0.021
ORB	0.0140	0.003	5.316	0.000	0.009	0.019
STL	0.0018	0.005	0.382	0.702	-0.007	0.011
BLK	0.0161	0.004	3.630	0.000	0.007	0.025
PF	-0.0095	0.002	-3.821	0.000	-0.014	-0.005
TS	-0.0240	0.015	-1.614	0.107	-0.053	0.005
X3PAR	-0.0109	0.006	-1.895	0.058	-0.022	0.000
ASTP	-6.748e-05	0.000	-0.489	0.625	-0.000	0.000
STLP	-0.0038	0.002	-2.306	0.021	-0.007	-0.001
BLKP	-0.0024	0.001	-3.013	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.139	0.033	3.26e-05	0.001

USG	0.0009	0.000	3.352	0.001	0.000	0.001
OWS	0.0024	0.001	3.762	0.000	0.001	0.004
DWS	-0.0021	0.001	-1.897	0.058	-0.004	6.94e-05
DBPM	0.0019	0.001	1.895	0.058	-6.6e-05	0.004

Omnibus:	4361.123	Durbin-Watson:	1.954
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4782605.590
Skew:	13.267	Prob(JB):	0.00
Kurtosis:	220.536	Cond. No.	577.

Warnings:

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

```
[96]: #removed STL not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'BLK',
         'PF', 'TS', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	MIP	R-squared:	0.071
Model:	OLS	Adj. R-squared:	0.063
Method:	Least Squares	F-statistic:	9.523
Date:	Wed, 12 May 2021	Prob (F-statistic):	3.00e-27
Time:	19:19:06	Log-Likelihood:	4147.8
No. Observations:	2390	AIC:	-8256.
Df Residuals:	2370	BIC:	-8140.
Df Model:	19		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0054	0.001	6.041	0.000	0.004	0.007
G	-7.135e-05	5.51e-05	-1.295	0.195	-0.000	3.67e-05
GS	8.759e-05	4.39e-05	1.994	0.046	1.46e-06	0.000

X3PM	0.0144	0.003	5.199	0.000	0.009	0.020
X3PP	-0.0025	0.006	-0.403	0.687	-0.015	0.010
X2PP	0.0148	0.017	0.883	0.377	-0.018	0.048
FTP	0.0053	0.008	0.648	0.517	-0.011	0.021
ORB	0.0141	0.003	5.390	0.000	0.009	0.019
BLK	0.0162	0.004	3.644	0.000	0.007	0.025
PF	-0.0092	0.002	-3.849	0.000	-0.014	-0.005
TS	-0.0237	0.015	-1.597	0.110	-0.053	0.005
X3PAR	-0.0110	0.006	-1.916	0.056	-0.022	0.000
ASTP	-6.639e-05	0.000	-0.481	0.631	-0.000	0.000
STLP	-0.0036	0.002	-2.346	0.019	-0.007	-0.001
BLKP	-0.0024	0.001	-2.993	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.113	0.035	2.78e-05	0.001
USG	0.0009	0.000	3.340	0.001	0.000	0.001
OWS	0.0024	0.001	3.746	0.000	0.001	0.004
DWS	-0.0020	0.001	-1.863	0.063	-0.004	0.000
DBPM	0.0018	0.001	1.863	0.063	-9.65e-05	0.004

Omnibus:	4360.960	Durbin-Watson:	1.954
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4782892.258
Skew:	13.266	Prob(JB):	0.00
Kurtosis:	220.543	Cond. No.	577.

Warnings:

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

```
[97]: #removed X3PP not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'ORB', 'BLK',
         'PF', 'TS', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	MIP	R-squared:	0.071
Model:	OLS	Adj. R-squared:	0.064

```

Method:                Least Squares      F-statistic:           10.05
Date:                  Wed, 12 May 2021    Prob (F-statistic):    9.98e-28
Time:                  19:19:06           Log-Likelihood:        4147.8
No. Observations:      2390              AIC:                   -8258.
Df Residuals:          2371              BIC:                   -8148.
Df Model:              18
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0054	0.001	6.038	0.000	0.004	0.007
G	-7.186e-05	5.51e-05	-1.305	0.192	-0.000	3.61e-05
GS	8.789e-05	4.39e-05	2.002	0.045	1.79e-06	0.000
X3PM	0.0142	0.003	5.218	0.000	0.009	0.020
X2PP	0.0150	0.017	0.896	0.370	-0.018	0.048
FTP	0.0052	0.008	0.630	0.529	-0.011	0.021
ORB	0.0142	0.003	5.400	0.000	0.009	0.019
BLK	0.0161	0.004	3.635	0.000	0.007	0.025
PF	-0.0092	0.002	-3.841	0.000	-0.014	-0.005
TS	-0.0244	0.015	-1.656	0.098	-0.053	0.005
X3PAR	-0.0110	0.006	-1.925	0.054	-0.022	0.000
ASTP	-6.499e-05	0.000	-0.471	0.638	-0.000	0.000
STLP	-0.0036	0.002	-2.361	0.018	-0.007	-0.001
BLKP	-0.0024	0.001	-2.992	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.106	0.035	2.65e-05	0.001
USG	0.0009	0.000	3.349	0.001	0.000	0.001
OWS	0.0024	0.001	3.755	0.000	0.001	0.004
DWS	-0.0020	0.001	-1.863	0.063	-0.004	0.000
DBPM	0.0019	0.001	1.891	0.059	-6.89e-05	0.004
Omnibus:	4360.867		Durbin-Watson:	1.954		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	4781927.099		
Skew:	13.266		Prob(JB):	0.00		
Kurtosis:	220.521		Cond. No.	576.		

Warnings:

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

```

[98]: #removed ASTP not statistically significant
cols = [ 'G', 'GS',
          'X3PM', 'X2PP',
          'FTP', 'ORB', 'BLK',
          'PF', 'TS', 'X3PAR',
          'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
          'DBPM', ]

```



```

x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())

```

OLS Regression Results

=====						
Dep. Variable:	MIP		R-squared:		0.071	
Model:	OLS		Adj. R-squared:		0.064	
Method:	Least Squares		F-statistic:		10.63	
Date:	Wed, 12 May 2021		Prob (F-statistic):		3.32e-28	
Time:	19:19:06		Log-Likelihood:		4147.6	
No. Observations:	2390		AIC:		-8259	
Df Residuals:	2372		BIC:		-8155.	
Df Model:	17					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.0054	0.001	6.029	0.000	0.004	0.007
G	-7.142e-05	5.51e-05	-1.297	0.195	-0.000	3.65e-05
GS	8.68e-05	4.38e-05	1.980	0.048	8.27e-07	0.000
X3PM	0.0142	0.003	5.228	0.000	0.009	0.020
X2PP	0.0153	0.017	0.914	0.361	-0.018	0.048
FTP	0.0051	0.008	0.621	0.535	-0.011	0.021
ORB	0.0142	0.003	5.427	0.000	0.009	0.019
BLK	0.0160	0.004	3.612	0.000	0.007	0.025
PF	-0.0093	0.002	-3.877	0.000	-0.014	-0.005
TS	-0.0236	0.015	-1.612	0.107	-0.052	0.005
X3PAR	-0.0113	0.006	-1.988	0.047	-0.023	-0.000
STLP	-0.0036	0.002	-2.355	0.019	-0.007	-0.001
BLKP	-0.0023	0.001	-2.962	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.070	0.039	1.86e-05	0.001
USG	0.0008	0.000	3.558	0.000	0.000	0.001
OWS	0.0024	0.001	3.728	0.000	0.001	0.004
DWS	-0.0019	0.001	-1.813	0.070	-0.004	0.000
DBPM	0.0017	0.001	1.834	0.067	-0.000	0.004
=====						
Omnibus:	4361.786		Durbin-Watson:		1.954	
Prob(Omnibus):	0.000		Jarque-Bera (JB):		4788595.447	
Skew:	13.271		Prob(JB):		0.00	
Kurtosis:	220.674		Cond. No.		575.	
=====						

Warnings:

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

```
[99]: #removed FTP not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'ORB', 'BLK',
         'PF', 'TS', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	MIP	R-squared:	0.071
Model:	OLS	Adj. R-squared:	0.064
Method:	Least Squares	F-statistic:	11.27
Date:	Wed, 12 May 2021	Prob (F-statistic):	1.15e-28
Time:	19:19:06	Log-Likelihood:	4147.5
No. Observations:	2390	AIC:	-8261.
Df Residuals:	2373	BIC:	-8163.
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0054	0.001	6.030	0.000	0.004	0.007
G	-6.906e-05	5.49e-05	-1.258	0.209	-0.000	3.86e-05
GS	8.657e-05	4.38e-05	1.975	0.048	6.19e-07	0.000
X3PM	0.0143	0.003	5.261	0.000	0.009	0.020
X2PP	0.0147	0.017	0.879	0.380	-0.018	0.047
ORB	0.0143	0.003	5.461	0.000	0.009	0.019
BLK	0.0161	0.004	3.633	0.000	0.007	0.025
PF	-0.0092	0.002	-3.858	0.000	-0.014	-0.005
TS	-0.0224	0.015	-1.543	0.123	-0.051	0.006
X3PAR	-0.0114	0.006	-2.005	0.045	-0.023	-0.000
STLP	-0.0036	0.002	-2.357	0.019	-0.007	-0.001
BLKP	-0.0023	0.001	-2.981	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.080	0.038	2.04e-05	0.001

USG	0.0008	0.000	3.589	0.000	0.000	0.001
OWS	0.0024	0.001	3.738	0.000	0.001	0.004
DWS	-0.0019	0.001	-1.835	0.067	-0.004	0.000
DBPM	0.0017	0.001	1.854	0.064	-0.000	0.004

Omnibus:	4362.239	Durbin-Watson:	1.954
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4792217.802
Skew:	13.274	Prob(JB):	0.00
Kurtosis:	220.756	Cond. No.	571.

Warnings:

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

```
[100]: #removed X2PP not statistically significant
cols = [ 'G', 'GS',
         'X3PM',
         'ORB', 'BLK',
         'PF', 'TS', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	MIP	R-squared:	0.070
Model:	OLS	Adj. R-squared:	0.064
Method:	Least Squares	F-statistic:	11.97
Date:	Wed, 12 May 2021	Prob (F-statistic):	4.66e-29
Time:	19:19:06	Log-Likelihood:	4147.1
No. Observations:	2390	AIC:	-8262.
Df Residuals:	2374	BIC:	-8170.
Df Model:	15		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0054	0.001	6.053	0.000	0.004	0.007
G	-6.433e-05	5.46e-05	-1.177	0.239	-0.000	4.28e-05
GS	8.67e-05	4.38e-05	1.978	0.048	7.52e-07	0.000

X3PM	0.0141	0.003	5.212	0.000	0.009	0.019
ORB	0.0143	0.003	5.474	0.000	0.009	0.019
BLK	0.0164	0.004	3.710	0.000	0.008	0.025
PF	-0.0092	0.002	-3.847	0.000	-0.014	-0.005
TS	-0.0184	0.014	-1.336	0.182	-0.045	0.009
X3PAR	-0.0113	0.006	-1.985	0.047	-0.022	-0.000
STLP	-0.0036	0.002	-2.347	0.019	-0.007	-0.001
BLKP	-0.0023	0.001	-2.999	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.079	0.038	2.03e-05	0.001
USG	0.0008	0.000	3.561	0.000	0.000	0.001
OWS	0.0024	0.001	3.828	0.000	0.001	0.004
DWS	-0.0020	0.001	-1.897	0.058	-0.004	6.79e-05
DBPM	0.0017	0.001	1.841	0.066	-0.000	0.004

Omnibus:	4362.208	Durbin-Watson:	1.955
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4790921.468
Skew:	13.274	Prob(JB):	0.00
Kurtosis:	220.727	Cond. No.	436.

Warnings:

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

```
[101]: #removed G not statistically significant
cols = [ 'GS',
         'X3PM',
         'ORB', 'BLK',
         'PF', 'TS', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	MIP	R-squared:	0.070
Model:	OLS	Adj. R-squared:	0.064
Method:	Least Squares	F-statistic:	12.73
Date:	Wed, 12 May 2021	Prob (F-statistic):	2.42e-29
Time:	19:19:06	Log-Likelihood:	4146.4
No. Observations:	2390	AIC:	-8263.

Df Residuals: 2375 BIC: -8176.
Df Model: 14
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.136	0.000	0.004	0.007
GS	7.397e-05	4.25e-05	1.741	0.082	-9.33e-06	0.000
X3PM	0.0140	0.003	5.168	0.000	0.009	0.019
ORB	0.0144	0.003	5.500	0.000	0.009	0.020
BLK	0.0163	0.004	3.685	0.000	0.008	0.025
PF	-0.0094	0.002	-3.955	0.000	-0.014	-0.005
TS	-0.0184	0.014	-1.333	0.183	-0.045	0.009
X3PAR	-0.0114	0.006	-2.008	0.045	-0.023	-0.000
STLP	-0.0036	0.002	-2.381	0.017	-0.007	-0.001
BLKP	-0.0023	0.001	-3.046	0.002	-0.004	-0.001
TOVP	0.0004	0.000	2.098	0.036	2.35e-05	0.001
USG	0.0008	0.000	3.720	0.000	0.000	0.001
OWS	0.0024	0.001	3.752	0.000	0.001	0.004
DWS	-0.0024	0.001	-2.328	0.020	-0.004	-0.000
DBPM	0.0019	0.001	1.971	0.049	9.87e-06	0.004
Omnibus:	4361.781	Durbin-Watson:	1.952			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4785748.987			
Skew:	13.272	Prob(JB):	0.00			
Kurtosis:	220.608	Cond. No.	399.			

Warnings:

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

```
[102]: #removed TS not statistically significant
cols = [ 'GS',
         'X3PM',
         'ORB', 'BLK',
         'PF', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```

=====
Dep. Variable:          MIP    R-squared:                0.069
Model:                  OLS    Adj. R-squared:            0.064
Method:                 Least Squares    F-statistic:          13.56
Date:                   Wed, 12 May 2021    Prob (F-statistic):    1.47e-29
Time:                   19:19:06    Log-Likelihood:        4145.5
No. Observations:       2390    AIC:                   -8263.
Df Residuals:           2376    BIC:                   -8182.
Df Model:               13
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.160	0.000	0.004	0.007
GS	7.699e-05	4.24e-05	1.815	0.070	-6.21e-06	0.000
X3PM	0.0137	0.003	5.067	0.000	0.008	0.019
ORB	0.0144	0.003	5.490	0.000	0.009	0.019
BLK	0.0163	0.004	3.685	0.000	0.008	0.025
PF	-0.0096	0.002	-4.045	0.000	-0.014	-0.005
X3PAR	-0.0117	0.006	-2.047	0.041	-0.023	-0.000
STLP	-0.0032	0.001	-2.136	0.033	-0.006	-0.000
BLKP	-0.0023	0.001	-2.989	0.003	-0.004	-0.001
TOVP	0.0004	0.000	2.057	0.040	1.65e-05	0.001
USG	0.0008	0.000	3.746	0.000	0.000	0.001
OWS	0.0020	0.001	3.522	0.000	0.001	0.003
DWS	-0.0022	0.001	-2.210	0.027	-0.004	-0.000
DBPM	0.0014	0.001	1.626	0.104	-0.000	0.003

```

=====
Omnibus:                 4362.876    Durbin-Watson:                1.951
Prob(Omnibus):            0.000    Jarque-Bera (JB):              4791400.123
Skew:                     13.279    Prob(JB):                      0.00
Kurtosis:                 220.737    Cond. No.                      170.
=====

```

Warnings:

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

```

[103]: #removed DBPM not statistically significant
cols = [ 'GS',
         'X3PM',
         'ORB', 'BLK',
         'PF', 'X3PAR',
         'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         ]
x_train = trainmip[cols] #all of the variables

```

```

y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          MIP    R-squared:                0.068
Model:                  OLS    Adj. R-squared:           0.063
Method:                 Least Squares    F-statistic:           14.46
Date:                  Wed, 12 May 2021    Prob (F-statistic):      1.30e-29
Time:                  19:19:06    Log-Likelihood:          4144.1
No. Observations:      2390    AIC:                     -8262.
Df Residuals:          2377    BIC:                     -8187.
Df Model:              12
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.150	0.000	0.004	0.007
GS	6.811e-05	4.21e-05	1.618	0.106	-1.44e-05	0.000
X3PM	0.0135	0.003	4.997	0.000	0.008	0.019
ORB	0.0140	0.003	5.358	0.000	0.009	0.019
BLK	0.0163	0.004	3.693	0.000	0.008	0.025
PF	-0.0094	0.002	-3.956	0.000	-0.014	-0.005
X3PAR	-0.0104	0.006	-1.835	0.067	-0.021	0.001
STLP	-0.0019	0.001	-1.498	0.134	-0.004	0.001
BLKP	-0.0017	0.001	-2.514	0.012	-0.003	-0.000
TOVP	0.0003	0.000	1.896	0.058	-1.1e-05	0.001
USG	0.0007	0.000	3.375	0.001	0.000	0.001
OWS	0.0020	0.001	3.592	0.000	0.001	0.003
DWS	-0.0016	0.001	-1.732	0.083	-0.003	0.000

```

=====
Omnibus:                4365.336    Durbin-Watson:           1.949
Prob(Omnibus):           0.000    Jarque-Bera (JB):        4805743.639
Skew:                    13.294    Prob(JB):                0.00
Kurtosis:                221.063    Cond. No.:               169.
=====

```

Warnings:

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

```
[104]: #removed STLP not statistically significant
cols = [ 'GS',
        'X3PM',
        'ORB', 'BLK',
        'PF', 'X3PAR',
        'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
        ]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MIP    R-squared:                0.067
Model:                  OLS    Adj. R-squared:           0.063
Method:                 Least Squares    F-statistic:           15.57
Date:                  Wed, 12 May 2021    Prob (F-statistic):      9.20e-30
Time:                  19:19:06    Log-Likelihood:         4143.0
No. Observations:      2390    AIC:                   -8262.
Df Residuals:          2378    BIC:                   -8193.
Df Model:               11
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.117	0.000	0.004	0.007
GS	7.028e-05	4.21e-05	1.670	0.095	-1.22e-05	0.000
X3PM	0.0135	0.003	5.002	0.000	0.008	0.019
ORB	0.0140	0.003	5.356	0.000	0.009	0.019
BLK	0.0159	0.004	3.610	0.000	0.007	0.025
PF	-0.0094	0.002	-3.944	0.000	-0.014	-0.005
X3PAR	-0.0103	0.006	-1.825	0.068	-0.021	0.001
BLKP	-0.0016	0.001	-2.365	0.018	-0.003	-0.000
TOVP	0.0003	0.000	1.848	0.065	-1.93e-05	0.001
USG	0.0007	0.000	3.336	0.001	0.000	0.001
OWS	0.0021	0.001	3.737	0.000	0.001	0.003
DWS	-0.0018	0.001	-1.990	0.047	-0.004	-2.69e-05

```
=====
Omnibus:                4365.733    Durbin-Watson:           1.951
Prob(Omnibus):           0.000    Jarque-Bera (JB):        4802558.752
Skew:                    13.298    Prob(JB):                 0.00
Kurtosis:                220.989    Cond. No.:                169.
=====
```


=====

Warnings:

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

```
[105]: #removed GS not statistically significant
cols = [
    'X3PM',
    'ORB', 'BLK',
    'PF', 'X3PAR',
    'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MIP    R-squared:                0.066
Model:                  OLS    Adj. R-squared:           0.062
Method:                 Least Squares    F-statistic:         16.83
Date:                   Wed, 12 May 2021    Prob (F-statistic):    8.15e-30
Time:                   19:19:06    Log-Likelihood:       4141.6
No. Observations:      2390    AIC:                 -8261.
Df Residuals:          2379    BIC:                 -8198.
Df Model:              10
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0055	0.001	6.087	0.000	0.004	0.007
X3PM	0.0145	0.003	5.480	0.000	0.009	0.020
ORB	0.0148	0.003	5.766	0.000	0.010	0.020
BLK	0.0162	0.004	3.669	0.000	0.008	0.025
PF	-0.0086	0.002	-3.689	0.000	-0.013	-0.004
X3PAR	-0.0105	0.006	-1.854	0.064	-0.022	0.001
BLKP	-0.0017	0.001	-2.476	0.013	-0.003	-0.000
TOVP	0.0003	0.000	1.883	0.060	-1.32e-05	0.001
USG	0.0007	0.000	3.210	0.001	0.000	0.001
OWS	0.0022	0.001	3.906	0.000	0.001	0.003
DWS	-0.0016	0.001	-1.785	0.074	-0.003	0.000

=====

Omnibus:	4366.883	Durbin-Watson:	1.950
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4806436.961
Skew:	13.305	Prob(JB):	0.00
Kurtosis:	221.076	Cond. No.	37.4

=====

Warnings:

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

```
[106]: #removed DWS not statistically significant
cols = [
    'X3PM',
    'ORB', 'BLK',
    'PF', 'X3PAR',
    'BLKP', 'TOVP', 'USG', 'OWS',
]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MIP    R-squared:                0.065
Model:                  OLS    Adj. R-squared:           0.061
Method:                 Least Squares    F-statistic:              18.33
Date:                   Wed, 12 May 2021    Prob (F-statistic):       8.38e-30
Time:                   19:19:07    Log-Likelihood:           4140.0
No. Observations:       2390    AIC:                      -8260.
Df Residuals:           2380    BIC:                      -8202.
Df Model:                9
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0054	0.001	6.054	0.000	0.004	0.007
X3PM	0.0145	0.003	5.507	0.000	0.009	0.020
ORB	0.0150	0.003	5.876	0.000	0.010	0.020
BLK	0.0151	0.004	3.450	0.001	0.006	0.024
PF	-0.0091	0.002	-3.910	0.000	-0.014	-0.005
X3PAR	-0.0095	0.006	-1.691	0.091	-0.021	0.002
BLKP	-0.0017	0.001	-2.578	0.010	-0.003	-0.000
TOVP	0.0003	0.000	1.849	0.065	-1.91e-05	0.001

USG	0.0006	0.000	3.160	0.002	0.000	0.001
OWS	0.0017	0.000	3.477	0.001	0.001	0.003

```
=====
```

Omnibus:	4371.799	Durbin-Watson:	1.949
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4837827.821
Skew:	13.336	Prob(JB):	0.00
Kurtosis:	221.790	Cond. No.	37.1

```
=====
```

Warnings:

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

```
[107]: #removed X3PAR not statistically significant
cols = [
    'X3PM',
    'ORB', 'BLK',
    'PF',
    'BLKP', 'TOVP', 'USG', 'OWS',
]
x_train = trainmip[cols] #all of the variables
y_train = trainmip['MIP']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
```

Dep. Variable:	MIP	R-squared:	0.064
Model:	OLS	Adj. R-squared:	0.061
Method:	Least Squares	F-statistic:	20.25
Date:	Wed, 12 May 2021	Prob (F-statistic):	7.05e-30
Time:	19:19:07	Log-Likelihood:	4138.6
No. Observations:	2390	AIC:	-8259.
Df Residuals:	2381	BIC:	-8207.
Df Model:	8		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.0053	0.001	5.947	0.000	0.004	0.007
X3PM	0.0135	0.003	5.253	0.000	0.008	0.018
ORB	0.0151	0.003	5.916	0.000	0.010	0.020
BLK	0.0138	0.004	3.212	0.001	0.005	0.022
PF	-0.0086	0.002	-3.724	0.000	-0.013	-0.004

BLKP	-0.0012	0.001	-2.061	0.039	-0.002	-6.04e-05
TOVP	0.0004	0.000	2.190	0.029	3.84e-05	0.001
USG	0.0007	0.000	3.353	0.001	0.000	0.001
OWS	0.0018	0.000	3.736	0.000	0.001	0.003

```
=====
Omnibus:                4374.607    Durbin-Watson:                1.949
Prob(Omnibus):           0.000    Jarque-Bera (JB):          4852958.569
Skew:                    13.355    Prob(JB):                  0.00
Kurtosis:                222.133    Cond. No.                  28.4
=====
```

Warnings:

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

```
[108]: import statsmodels.formula.api as smf

MIPLinear = smf.ols(formula='MIP ~ X3PM + ORB + BLK + PF + BLKP + TOVP + USG + OWS',
                    data=trainmip).fit()
print(MIPLinear.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          MIP    R-squared:                0.064
Model:                  OLS    Adj. R-squared:          0.061
Method:                 Least Squares    F-statistic:            20.25
Date:                   Wed, 12 May 2021    Prob (F-statistic):      7.05e-30
Time:                   19:19:07    Log-Likelihood:          4138.6
No. Observations:       2390    AIC:                     -8259.
Df Residuals:           2381    BIC:                     -8207.
Df Model:                8
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.0053	0.001	5.947	0.000	0.004	0.007
X3PM	0.0135	0.003	5.253	0.000	0.008	0.018
ORB	0.0151	0.003	5.916	0.000	0.010	0.020
BLK	0.0138	0.004	3.212	0.001	0.005	0.022
PF	-0.0086	0.002	-3.724	0.000	-0.013	-0.004
BLKP	-0.0012	0.001	-2.061	0.039	-0.002	-6.04e-05
TOVP	0.0004	0.000	2.190	0.029	3.84e-05	0.001
USG	0.0007	0.000	3.353	0.001	0.000	0.001
OWS	0.0018	0.000	3.736	0.000	0.001	0.003

```
=====
Omnibus:                4374.607    Durbin-Watson:                1.949
Prob(Omnibus):           0.000    Jarque-Bera (JB):          4852958.569
```

```
Skew:                13.355    Prob(JB):                0.00
Kurtosis:            222.133    Cond. No.                28.4
=====
```

Warnings:

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

```
[109]: OSR2(MIPLinear, trainmip, testmip, 'MIP')
```

```
[109]: 0.03547910086759887
```

Rookie of the Year

```
[185]: #loading in ROY data
datasetroy = pd.read_csv('roydata142.csv')
trainroy = datasetroy[datasetroy['Season'] <= 2017]
testroy = datasetroy[datasetroy['Season'] > 2017]
testroy
```

```
[185]:      Unnamed: 0      Player  Season Pos  Age  Tm  G  GS  MP  FGM  \
4          5      Aaron Holiday    2019  PG   22  IND  50   0  12.9  2.1
5          6      Abdel Nader    2018  SF   24  BOS  48   1  10.9  1.0
6          7  Admiral Schofield    2020  PF   22  WAS  33   2  11.2  1.1
11         12      Alex Caruso    2018  PG   23  LAL  37   7  15.2  1.4
14         15      Allonzo Trier    2019  SG   23  NYK  64   3  22.8  3.6
..         ...      ...      ...  ..  ...  ..  ..  ..  ..  ..
555        556  Vincent Poirier    2020   C   26  BOS  22   0   5.9  0.8
559        560  Wenyen Gabriel    2020  PF   22  TOT  30   1   7.8  0.7
560        561  Wesley Iwundu    2018  SF   23  ORL  62  12  16.5  1.5
570        571    Zach Collins    2018   C   20  POR  66   1  15.8  1.7
572        573  Zion Williamson    2020  PF   19  NOP  24  24  27.8  8.8

      ...  USG  OWS  DWS  WS  WS48  OBPM  DBPM  BPM  VORP  ROY
4      ...  21.9  0.1  0.8  0.9  0.065  -1.7   0.1  -1.6   0.1  0.00
5      ...  17.1 -0.9  0.8 -0.1 -0.014  -5.5   0.2  -5.3  -0.4  0.00
6      ...  11.9  0.1  0.1  0.1  0.017  -4.3  -1.2  -5.5  -0.3  0.00
11     ...  12.9 -0.1  0.6  0.5  0.041  -3.8   0.9  -2.9  -0.1  0.00
14     ...  21.5  0.4  0.5  0.9  0.030  -1.6  -1.5  -3.1  -0.4  0.00
..     ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...
555    ...  15.4  0.1  0.2  0.3  0.113  -4.8   1.0  -3.8  -0.1  0.00
559    ...  13.4  0.0  0.2  0.2  0.048  -3.7  -0.1  -3.8  -0.1  0.00
560    ...  11.6  0.0  0.7  0.7  0.032  -4.2  -0.1  -4.3  -0.6  0.00
570    ...  15.4 -0.7  1.3  0.6  0.030  -4.5   0.2  -4.3  -0.6  0.00
572    ...  30.5  1.5  0.4  2.0  0.141   3.5  -1.4  2.1   0.7  0.28
```

[157 rows x 52 columns]

```
[111]: #removed age b/c not relevant to predicting our output variable
cols = [ 'G', 'GS',
         'X3PM', 'X3PP', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
```

Dep. Variable:	ROY	R-squared:	0.434
Model:	OLS	Adj. R-squared:	0.403
Method:	Least Squares	F-statistic:	13.71
Date:	Wed, 12 May 2021	Prob (F-statistic):	1.96e-36
Time:	19:19:07	Log-Likelihood:	321.54
No. Observations:	416	AIC:	-597.1
Df Residuals:	393	BIC:	-504.4
Df Model:	22		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-0.0296	0.087	-0.342	0.733	-0.200	0.141
G	-0.0008	0.000	-1.840	0.066	-0.002	5.62e-05
GS	0.0020	0.000	4.580	0.000	0.001	0.003
X3PM	0.0407	0.026	1.580	0.115	-0.010	0.091
X3PP	0.0079	0.051	0.155	0.877	-0.092	0.107
X2PP	0.3159	0.126	2.505	0.013	0.068	0.564
FTP	0.1416	0.055	2.586	0.010	0.034	0.249
ORB	-0.0074	0.021	-0.360	0.719	-0.048	0.033
STL	0.1064	0.031	3.387	0.001	0.045	0.168
BLK	0.0862	0.037	2.342	0.020	0.014	0.159
PF	-0.0185	0.015	-1.224	0.222	-0.048	0.011
TS	-0.7847	0.161	-4.875	0.000	-1.101	-0.468
FTR	0.1299	0.044	2.936	0.004	0.043	0.217
DRBP	0.0029	0.002	1.691	0.092	-0.000	0.006
X3PAR	-0.0396	0.053	-0.752	0.453	-0.143	0.064
ASTP	0.0015	0.001	1.345	0.179	-0.001	0.004
STLP	-0.0383	0.013	-2.910	0.004	-0.064	-0.012

BLKP	-0.0128	0.008	-1.611	0.108	-0.029	0.003
TOVP	0.0014	0.002	0.892	0.373	-0.002	0.004
USG	0.0075	0.002	3.920	0.000	0.004	0.011
OWS	0.0310	0.007	4.570	0.000	0.018	0.044
DWS	-0.0088	0.014	-0.632	0.528	-0.036	0.019
DBPM	0.0194	0.010	1.949	0.052	-0.000	0.039

```
=====
Omnibus:                278.849    Durbin-Watson:                2.139
Prob(Omnibus):           0.000    Jarque-Bera (JB):            3489.838
Skew:                    2.719    Prob(JB):                     0.00
Kurtosis:                16.106    Cond. No.                     2.19e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.19e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[112]: #removed X3PP b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'ORB', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            ROY    R-squared:                0.434
Model:                    OLS    Adj. R-squared:           0.404
Method:                    Least Squares    F-statistic:              14.40
Date:                      Wed, 12 May 2021    Prob (F-statistic):       5.13e-37
Time:                      19:19:07    Log-Likelihood:           321.53
No. Observations:          416    AIC:                      -599.1
Df Residuals:              394    BIC:                      -510.4
Df Model:                  21
Covariance Type:           nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

```

-----
const      -0.0296      0.087      -0.341      0.733      -0.200      0.141
G           -0.0008      0.000      -1.836      0.067      -0.002      5.78e-05
GS          0.0020      0.000      4.588      0.000      0.001      0.003
X3PM        0.0416      0.025      1.654      0.099      -0.008      0.091
X2PP        0.3120      0.123      2.527      0.012      0.069      0.555
FTP         0.1416      0.055      2.589      0.010      0.034      0.249
ORB        -0.0077      0.020      -0.378      0.706      -0.048      0.032
STL         0.1064      0.031      3.395      0.001      0.045      0.168
BLK         0.0864      0.037      2.350      0.019      0.014      0.159
PF          -0.0187      0.015      -1.239      0.216      -0.048      0.011
TS          -0.7807      0.159      -4.919      0.000      -1.093      -0.469
FTR         0.1304      0.044      2.959      0.003      0.044      0.217
DRBP        0.0030      0.002      1.723      0.086      -0.000      0.006
X3PAR       -0.0379      0.051      -0.736      0.462      -0.139      0.063
ASTP        0.0016      0.001      1.361      0.174      -0.001      0.004
STLP        -0.0381      0.013      -2.911      0.004      -0.064      -0.012
BLKP        -0.0129      0.008      -1.627      0.105      -0.029      0.003
TOVP        0.0014      0.002      0.891      0.374      -0.002      0.004
USG         0.0075      0.002      3.928      0.000      0.004      0.011
OWS         0.0310      0.007      4.575      0.000      0.018      0.044
DWS         -0.0088      0.014      -0.633      0.527      -0.036      0.019
DBPM        0.0194      0.010      1.949      0.052      -0.000      0.039
=====
Omnibus:                279.009   Durbin-Watson:                2.139
Prob(Omnibus):           0.000   Jarque-Bera (JB):            3498.433
Skew:                    2.720   Prob(JB):                     0.00
Kurtosis:                16.124   Cond. No.                     2.15e+03
=====

```

Warnings:

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

[2] The condition number is large, 2.15e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```

[113]: #removed ORB b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS', 'DWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

```



```
#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY      R-squared:          0.434
Model:                  OLS      Adj. R-squared:       0.405
Method:                 Least Squares      F-statistic:       15.14
Date:                  Wed, 12 May 2021      Prob (F-statistic):    1.39e-37
Time:                  19:19:07      Log-Likelihood:       321.45
No. Observations:      416      AIC:                -600.9
Df Residuals:          395      BIC:                -516.3
Df Model:               20
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0262	0.086	-0.304	0.761	-0.196	0.143
G	-0.0008	0.000	-1.839	0.067	-0.002	5.66e-05
GS	0.0020	0.000	4.577	0.000	0.001	0.003
X3PM	0.0449	0.024	1.909	0.057	-0.001	0.091
X2PP	0.3034	0.121	2.503	0.013	0.065	0.542
FTP	0.1447	0.054	2.677	0.008	0.038	0.251
STL	0.1061	0.031	3.389	0.001	0.045	0.168
BLK	0.0809	0.034	2.399	0.017	0.015	0.147
PF	-0.0210	0.014	-1.532	0.126	-0.048	0.006
TS	-0.7779	0.158	-4.912	0.000	-1.089	-0.467
FTR	0.1300	0.044	2.953	0.003	0.043	0.216
DRBP	0.0028	0.002	1.696	0.091	-0.000	0.006
X3PAR	-0.0384	0.051	-0.746	0.456	-0.139	0.063
ASTP	0.0016	0.001	1.452	0.147	-0.001	0.004
STLP	-0.0393	0.013	-3.078	0.002	-0.064	-0.014
BLKP	-0.0123	0.008	-1.585	0.114	-0.028	0.003
TOVP	0.0014	0.002	0.889	0.375	-0.002	0.004
USG	0.0075	0.002	3.933	0.000	0.004	0.011
OWS	0.0302	0.006	4.697	0.000	0.018	0.043
DWS	-0.0094	0.014	-0.676	0.499	-0.037	0.018
DBPM	0.0207	0.009	2.215	0.027	0.002	0.039

```
=====
Omnibus:                278.459      Durbin-Watson:         2.135
Prob(Omnibus):          0.000      Jarque-Bera (JB):      3474.206
Skew:                   2.715      Prob(JB):              0.00
Kurtosis:               16.075      Cond. No.              2.14e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.14e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[114]: #removed DWS b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP', 'X3PAR',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY      R-squared:                0.433
Model:                  OLS      Adj. R-squared:            0.406
Method:                 Least Squares      F-statistic:        15.94
Date:                   Wed, 12 May 2021    Prob (F-statistic):    4.23e-38
Time:                   19:19:07    Log-Likelihood:        321.21
No. Observations:      416      AIC:                  -602.4
Df Residuals:          396      BIC:                  -521.8
Df Model:               19
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0265	0.086	-0.308	0.759	-0.196	0.143
G	-0.0010	0.000	-2.381	0.018	-0.002	-0.000
GS	0.0019	0.000	4.551	0.000	0.001	0.003
X3PM	0.0471	0.023	2.024	0.044	0.001	0.093
X2PP	0.3036	0.121	2.506	0.013	0.065	0.542
FTP	0.1461	0.054	2.707	0.007	0.040	0.252
STL	0.1011	0.030	3.326	0.001	0.041	0.161
BLK	0.0794	0.034	2.362	0.019	0.013	0.146
PF	-0.0219	0.014	-1.604	0.109	-0.049	0.005
TS	-0.7626	0.157	-4.869	0.000	-1.071	-0.455
FTR	0.1285	0.044	2.926	0.004	0.042	0.215
DRBP	0.0025	0.002	1.588	0.113	-0.001	0.006

X3PAR	-0.0418	0.051	-0.818	0.414	-0.142	0.059
ASTP	0.0017	0.001	1.502	0.134	-0.001	0.004
STLP	-0.0368	0.012	-3.011	0.003	-0.061	-0.013
BLKP	-0.0116	0.008	-1.506	0.133	-0.027	0.004
TOVP	0.0013	0.002	0.856	0.393	-0.002	0.004
USG	0.0072	0.002	3.897	0.000	0.004	0.011
OWS	0.0290	0.006	4.696	0.000	0.017	0.041
DBPM	0.0179	0.008	2.138	0.033	0.001	0.034

```
=====
Omnibus:                280.450    Durbin-Watson:                2.130
Prob(Omnibus):          0.000    Jarque-Bera (JB):          3562.150
Skew:                   2.735    Prob(JB):                  0.00
Kurtosis:               16.251    Cond. No.                  2.13e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.13e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[115]: #removed X3PAR b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'TOVP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY    R-squared:                0.432
Model:                  OLS    Adj. R-squared:          0.407
Method:                 Least Squares    F-statistic:            16.80
Date:                   Wed, 12 May 2021    Prob (F-statistic):      1.39e-38
Time:                   19:19:07    Log-Likelihood:          320.86
No. Observations:       416    AIC:                    -603.7
Df Residuals:           397    BIC:                    -527.1
Df Model:                18
Covariance Type:        nonrobust
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0600	0.076	-0.793	0.428	-0.209	0.089
G	-0.0009	0.000	-2.263	0.024	-0.002	-0.000
GS	0.0020	0.000	4.594	0.000	0.001	0.003
X3PM	0.0355	0.018	1.924	0.055	-0.001	0.072
X2PP	0.2956	0.121	2.449	0.015	0.058	0.533
FTP	0.1437	0.054	2.668	0.008	0.038	0.250
STL	0.1049	0.030	3.493	0.001	0.046	0.164
BLK	0.0731	0.033	2.235	0.026	0.009	0.137
PF	-0.0204	0.014	-1.510	0.132	-0.047	0.006
TS	-0.7456	0.155	-4.805	0.000	-1.051	-0.440
FTR	0.1354	0.043	3.143	0.002	0.051	0.220
DRBP	0.0028	0.002	1.780	0.076	-0.000	0.006
ASTP	0.0017	0.001	1.540	0.124	-0.000	0.004
STLP	-0.0359	0.012	-2.946	0.003	-0.060	-0.012
BLKP	-0.0091	0.007	-1.286	0.199	-0.023	0.005
TOVP	0.0014	0.002	0.910	0.363	-0.002	0.004
USG	0.0075	0.002	4.146	0.000	0.004	0.011
OWS	0.0290	0.006	4.710	0.000	0.017	0.041
DBPM	0.0160	0.008	1.992	0.047	0.000	0.032
=====						
Omnibus:		280.894	Durbin-Watson:		2.129	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		3586.684	
Skew:		2.738	Prob(JB):		0.00	
Kurtosis:		16.301	Cond. No.		2.11e+03	
=====						

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.11e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[116]: #removed TOVP b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'BLKP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
modell= sm.OLS(y_train, x_train).fit()
print(modell.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY      R-squared:                0.431
Model:                  OLS      Adj. R-squared:           0.407
Method:                 Least Squares      F-statistic:        17.75
Date:                   Wed, 12 May 2021    Prob (F-statistic):    4.79e-39
Time:                   19:19:07    Log-Likelihood:       320.43
No. Observations:      416      AIC:                 -604.9
Df Residuals:          398      BIC:                 -532.3
Df Model:               17
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0476	0.074	-0.639	0.523	-0.194	0.099
G	-0.0009	0.000	-2.265	0.024	-0.002	-0.000
GS	0.0020	0.000	4.632	0.000	0.001	0.003
X3PM	0.0344	0.018	1.868	0.062	-0.002	0.071
X2PP	0.2921	0.121	2.422	0.016	0.055	0.529
FTP	0.1397	0.054	2.604	0.010	0.034	0.245
STL	0.0999	0.030	3.384	0.001	0.042	0.158
BLK	0.0722	0.033	2.208	0.028	0.008	0.136
PF	-0.0184	0.013	-1.377	0.169	-0.045	0.008
TS	-0.7339	0.155	-4.746	0.000	-1.038	-0.430
FTR	0.1445	0.042	3.449	0.001	0.062	0.227
DRBP	0.0029	0.002	1.863	0.063	-0.000	0.006
ASTP	0.0023	0.001	2.402	0.017	0.000	0.004
STLP	-0.0344	0.012	-2.851	0.005	-0.058	-0.011
BLKP	-0.0087	0.007	-1.236	0.217	-0.023	0.005
USG	0.0071	0.002	4.046	0.000	0.004	0.011
OWS	0.0271	0.006	4.689	0.000	0.016	0.038
DBPM	0.0158	0.008	1.968	0.050	1.38e-05	0.032

```
=====
Omnibus:                281.383    Durbin-Watson:          2.119
Prob(Omnibus):           0.000    Jarque-Bera (JB):       3595.778
Skew:                    2.745    Prob(JB):               0.00
Kurtosis:               16.316    Cond. No.               2.06e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.06e+03. This might indicate that there are

strong multicollinearity or other numerical problems.

```
[117]: #removed BLKP b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'PF', 'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

Dep. Variable:	ROY	R-squared:	0.429
Model:	OLS	Adj. R-squared:	0.406
Method:	Least Squares	F-statistic:	18.74
Date:	Wed, 12 May 2021	Prob (F-statistic):	2.24e-39
Time:	19:19:07	Log-Likelihood:	319.63
No. Observations:	416	AIC:	-605.3
Df Residuals:	399	BIC:	-536.7
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0676	0.073	-0.930	0.353	-0.210	0.075
G	-0.0008	0.000	-2.140	0.033	-0.002	-6.77e-05
GS	0.0020	0.000	4.670	0.000	0.001	0.003
X3PM	0.0323	0.018	1.758	0.079	-0.004	0.068
X2PP	0.2893	0.121	2.398	0.017	0.052	0.527
FTP	0.1515	0.053	2.867	0.004	0.048	0.255
STL	0.1050	0.029	3.588	0.000	0.047	0.162
BLK	0.0465	0.025	1.840	0.067	-0.003	0.096
PF	-0.0160	0.013	-1.213	0.226	-0.042	0.010
TS	-0.7391	0.155	-4.779	0.000	-1.043	-0.435
FTR	0.1339	0.041	3.262	0.001	0.053	0.215
DRBP	0.0025	0.002	1.667	0.096	-0.000	0.006
ASTP	0.0023	0.001	2.443	0.015	0.000	0.004
STLP	-0.0314	0.012	-2.654	0.008	-0.055	-0.008
USG	0.0072	0.002	4.125	0.000	0.004	0.011
OWS	0.0274	0.006	4.752	0.000	0.016	0.039

DBPM	0.0129	0.008	1.679	0.094	-0.002	0.028
------	--------	-------	-------	-------	--------	-------

```
=====
```

Omnibus:	279.971	Durbin-Watson:	2.108
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3593.780
Skew:	2.723	Prob(JB):	0.00
Kurtosis:	16.329	Cond. No.	2.06e+03

```
=====
```

Warnings:

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

[2] The condition number is large, 2.06e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[118]: #removed PF b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL', 'BLK',
         'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
```

Dep. Variable:	ROY	R-squared:	0.427
Model:	OLS	Adj. R-squared:	0.405
Method:	Least Squares	F-statistic:	19.87
Date:	Wed, 12 May 2021	Prob (F-statistic):	9.92e-40
Time:	19:19:07	Log-Likelihood:	318.86
No. Observations:	416	AIC:	-605.7
Df Residuals:	400	BIC:	-541.2
Df Model:	15		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

```
-----
```

const	-0.0658	0.073	-0.905	0.366	-0.209	0.077
G	-0.0009	0.000	-2.468	0.014	-0.002	-0.000
GS	0.0019	0.000	4.539	0.000	0.001	0.003
X3PM	0.0321	0.018	1.746	0.082	-0.004	0.068

X2PP	0.2827	0.121	2.344	0.020	0.046	0.520
FTP	0.1482	0.053	2.806	0.005	0.044	0.252
STL	0.0926	0.027	3.375	0.001	0.039	0.147
BLK	0.0350	0.023	1.493	0.136	-0.011	0.081
TS	-0.7499	0.154	-4.854	0.000	-1.054	-0.446
FTR	0.1327	0.041	3.233	0.001	0.052	0.213
DRBP	0.0024	0.002	1.601	0.110	-0.001	0.005
ASTP	0.0024	0.001	2.546	0.011	0.001	0.004
STLP	-0.0296	0.012	-2.524	0.012	-0.053	-0.007
USG	0.0071	0.002	4.079	0.000	0.004	0.011
OWS	0.0279	0.006	4.830	0.000	0.017	0.039
DBPM	0.0126	0.008	1.632	0.103	-0.003	0.028

```
=====
Omnibus:                279.113    Durbin-Watson:                2.115
Prob(Omnibus):           0.000    Jarque-Bera (JB):            3565.398
Skew:                    2.713    Prob(JB):                     0.00
Kurtosis:               16.276    Cond. No.                     2.06e+03
=====
```

Warnings:

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

[2] The condition number is large, 2.06e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[119]: #removed BLK b/c not statistically significant
cols = [ 'G', 'GS',
         'X3PM', 'X2PP',
         'FTP', 'STL',
         'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
modell1= sm.OLS(y_train, x_train).fit()
print(modell1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY    R-squared:                0.424
Model:                  OLS    Adj. R-squared:           0.404
Method:                 Least Squares    F-statistic:            21.06
Date:                   Wed, 12 May 2021    Prob (F-statistic):      6.12e-40
Time:                   19:19:07    Log-Likelihood:          317.71
=====
```


No. Observations: 416 AIC: -605.4
Df Residuals: 401 BIC: -545.0
Df Model: 14
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0667	0.073	-0.916	0.360	-0.210	0.076
G	-0.0009	0.000	-2.457	0.014	-0.002	-0.000
GS	0.0021	0.000	5.086	0.000	0.001	0.003
X3PM	0.0254	0.018	1.422	0.156	-0.010	0.060
X2PP	0.3008	0.120	2.504	0.013	0.065	0.537
FTP	0.1445	0.053	2.735	0.007	0.041	0.248
STL	0.1020	0.027	3.814	0.000	0.049	0.155
TS	-0.7542	0.155	-4.875	0.000	-1.058	-0.450
FTR	0.1363	0.041	3.321	0.001	0.056	0.217
DRBP	0.0028	0.002	1.890	0.060	-0.000	0.006
ASTP	0.0022	0.001	2.313	0.021	0.000	0.004
STLP	-0.0361	0.011	-3.307	0.001	-0.058	-0.015
USG	0.0079	0.002	4.655	0.000	0.005	0.011
OWS	0.0273	0.006	4.740	0.000	0.016	0.039
DBPM	0.0172	0.007	2.442	0.015	0.003	0.031
Omnibus:	276.838	Durbin-Watson:	2.116			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3557.516			
Skew:	2.679	Prob(JB):	0.00			
Kurtosis:	16.287	Cond. No.	2.05e+03			

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.05e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[120]: #removed X3PM b/c not statistically significant
cols = [ 'G', 'GS',
         'X2PP',
         'FTP', 'STL',
         'TS', 'FTR', 'DRBP',
         'ASTP', 'STLP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)
```

```
#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY      R-squared:                0.421
Model:                  OLS      Adj. R-squared:           0.402
Method:                 Least Squares      F-statistic:        22.47
Date:                  Wed, 12 May 2021      Prob (F-statistic):    3.32e-40
Time:                  19:19:07      Log-Likelihood:       316.66
No. Observations:      416      AIC:                 -605.3
Df Residuals:          402      BIC:                 -548.9
Df Model:               13
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0763	0.073	-1.051	0.294	-0.219	0.066
G	-0.0008	0.000	-2.237	0.026	-0.002	-0.000
GS	0.0021	0.000	5.212	0.000	0.001	0.003
X2PP	0.2638	0.117	2.246	0.025	0.033	0.495
FTP	0.1574	0.052	3.021	0.003	0.055	0.260
STL	0.1112	0.026	4.281	0.000	0.060	0.162
TS	-0.7045	0.151	-4.669	0.000	-1.001	-0.408
FTR	0.1247	0.040	3.096	0.002	0.046	0.204
DRBP	0.0023	0.001	1.586	0.114	-0.001	0.005
ASTP	0.0021	0.001	2.280	0.023	0.000	0.004
STLP	-0.0385	0.011	-3.566	0.000	-0.060	-0.017
USG	0.0083	0.002	4.956	0.000	0.005	0.012
OWS	0.0273	0.006	4.729	0.000	0.016	0.039
DBPM	0.0174	0.007	2.462	0.014	0.004	0.031

```
=====
Omnibus:                279.881      Durbin-Watson:          2.132
Prob(Omnibus):          0.000      Jarque-Bera (JB):      3659.776
Skew:                   2.713      Prob(JB):              0.00
Kurtosis:               16.479      Cond. No.              1.99e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.99e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[121]: #removed DRBP b/c not statistically significant
cols = [ 'G', 'GS',
```

```

        'X2PP',
        'FTP', 'STL',
        'TS', 'FTR',
        'ASTP', 'STLP', 'USG', 'OWS',
        'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())

```

OLS Regression Results

Dep. Variable:	ROY	R-squared:	0.417
Model:	OLS	Adj. R-squared:	0.400
Method:	Least Squares	F-statistic:	24.04
Date:	Wed, 12 May 2021	Prob (F-statistic):	2.20e-40
Time:	19:19:07	Log-Likelihood:	315.36
No. Observations:	416	AIC:	-604.7
Df Residuals:	403	BIC:	-552.3
Df Model:	12		
Covariance Type:	nonrobust		
=====			
	coef	std err	t P> t [0.025 0.975]

const	-0.0402	0.069	-0.582 0.561 -0.176 0.096
G	-0.0008	0.000	-2.129 0.034 -0.002 -6.1e-05
GS	0.0021	0.000	5.168 0.000 0.001 0.003
X2PP	0.2848	0.117	2.436 0.015 0.055 0.515
FTP	0.1504	0.052	2.890 0.004 0.048 0.253
STL	0.1149	0.026	4.431 0.000 0.064 0.166
TS	-0.7323	0.150	-4.877 0.000 -1.027 -0.437
FTR	0.1390	0.039	3.535 0.000 0.062 0.216
ASTP	0.0016	0.001	1.820 0.069 -0.000 0.003
STLP	-0.0438	0.010	-4.248 0.000 -0.064 -0.024
USG	0.0090	0.002	5.588 0.000 0.006 0.012
OWS	0.0284	0.006	4.951 0.000 0.017 0.040
DBPM	0.0211	0.007	3.161 0.002 0.008 0.034
=====			
Omnibus:	280.871	Durbin-Watson:	2.122
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3665.526
Skew:	2.728	Prob(JB):	0.00
Kurtosis:	16.480	Cond. No.	1.93e+03
=====			

Warnings:

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

[2] The condition number is large, 1.93e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[122]: #removed ASTP b/c not statistically significant
cols = [ 'G', 'GS',
         'X2PP',
         'FTP', 'STL',
         'TS', 'FTR',
         'STLP', 'USG', 'OWS',
         'DBPM', ]
x_train = trainroy[cols] #all of the variables
y_train = trainroy['ROY']
#adding an intercept
x_train = sm.add_constant(x_train)

#fit data to the model
model1= sm.OLS(y_train, x_train).fit()
print(model1.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          ROY      R-squared:                0.412
Model:                  OLS      Adj. R-squared:           0.396
Method:                 Least Squares      F-statistic:       25.78
Date:                   Wed, 12 May 2021    Prob (F-statistic):   2.08e-40
Time:                   19:19:08           Log-Likelihood:    313.66
No. Observations:      416             AIC:                -603.3
Df Residuals:          404             BIC:                -555.0
Df Model:               11
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0361	0.069	-0.522	0.602	-0.172	0.100
G	-0.0008	0.000	-2.092	0.037	-0.002	-4.72e-05
GS	0.0020	0.000	4.939	0.000	0.001	0.003
X2PP	0.2985	0.117	2.551	0.011	0.068	0.529
FTP	0.1674	0.051	3.263	0.001	0.067	0.268
STL	0.1304	0.025	5.311	0.000	0.082	0.179
TS	-0.7867	0.148	-5.331	0.000	-1.077	-0.497
FTR	0.1336	0.039	3.398	0.001	0.056	0.211
STLP	-0.0422	0.010	-4.100	0.000	-0.062	-0.022
USG	0.0098	0.002	6.368	0.000	0.007	0.013
OWS	0.0289	0.006	5.021	0.000	0.018	0.040

DBPM	0.0200	0.007	2.996	0.003	0.007	0.033
------	--------	-------	-------	-------	-------	-------

```
=====
```

Omnibus:	286.165	Durbin-Watson:	2.125
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3887.467
Skew:	2.785	Prob(JB):	0.00
Kurtosis:	16.902	Cond. No.	1.87e+03

```
=====
```

Warnings:

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

[2] The condition number is large, 1.87e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[123]: import statsmodels.formula.api as smf

ROYLinear = smf.ols(formula='ROY ~ G + GS + FTP + X2PP+ STL + TS + FTR + STLPL +
↳USG + OWS + DBPM',
                    data=trainroy).fit()
print(ROYLinear.summary())
```

OLS Regression Results

```
=====
```

Dep. Variable:	ROY	R-squared:	0.412
Model:	OLS	Adj. R-squared:	0.396
Method:	Least Squares	F-statistic:	25.78
Date:	Wed, 12 May 2021	Prob (F-statistic):	2.08e-40
Time:	19:19:08	Log-Likelihood:	313.66
No. Observations:	416	AIC:	-603.3
Df Residuals:	404	BIC:	-555.0
Df Model:	11		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

```
-----
```

Intercept	-0.0361	0.069	-0.522	0.602	-0.172	0.100
G	-0.0008	0.000	-2.092	0.037	-0.002	-4.72e-05
GS	0.0020	0.000	4.939	0.000	0.001	0.003
FTP	0.1674	0.051	3.263	0.001	0.067	0.268
X2PP	0.2985	0.117	2.551	0.011	0.068	0.529
STL	0.1304	0.025	5.311	0.000	0.082	0.179
TS	-0.7867	0.148	-5.331	0.000	-1.077	-0.497
FTR	0.1336	0.039	3.398	0.001	0.056	0.211
STLP	-0.0422	0.010	-4.100	0.000	-0.062	-0.022
USG	0.0098	0.002	6.368	0.000	0.007	0.013
OWS	0.0289	0.006	5.021	0.000	0.018	0.040
DBPM	0.0200	0.007	2.996	0.003	0.007	0.033

```
=====
```

Omnibus:	286.165	Durbin-Watson:	2.125
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3887.467
Skew:	2.785	Prob(JB):	0.00
Kurtosis:	16.902	Cond. No.	1.87e+03

=====

Warnings:

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

[2] The condition number is large, 1.87e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[124]: OSR2(ROYLinear, trainroy, testroy, 'ROY')
```

```
[124]: 0.42183685623993505
```

2021 Predictions

```
[131]: #mvp 2021 prediction
nba2021 = pd.read_csv('nba2021.csv')
y_predicted = MVPLinear.predict(nba2021)
finallmvp = pd.concat([y_predicted,nba2021],axis=1)
finallmvp = finallmvp.rename(columns={0:'Predicted MVP'})
finallmvp = finallmvp.sort_values(by='Predicted MVP',ascending = False)
finallmvp = finallmvp.loc[finallmvp['GS'] > 30]
finallmvp['Player'].iloc[0]
```

```
[131]: 'Nikola Jokic'
```

```
[132]: finallmvp.head(5)
```

```
[132]:
```

	Predicted MVP	Unnamed: 0	Player	Season	Pos	Age	Tm	\
259	0.150505	260	Nikola Jokic	2021	C	25	DEN	
294	0.134981	295	Damian Lillard	2021	PG	30	POR	
114	0.124799	115	Stephen Curry	2021	PG	32	GSW	
11	0.105019	12	Giannis Antetokounmpo	2021	PF	26	MIL	
82	0.096915	83	Jimmy Butler	2021	SF	31	MIA	

	G	GS	MP	...	OWS	DWS	WS	WS48	OBPM	DBPM	BPM	VORP	MVP	\
259	69	69	35.0	...	11.7	3.3	15.0	0.298	9.1	2.5	11.5	8.2	0	
294	64	64	35.7	...	9.0	0.8	9.8	0.205	7.3	-1.6	5.7	4.5	0	
114	62	62	34.1	...	6.5	2.5	8.9	0.203	8.0	0.0	8.1	5.4	0	
11	59	59	32.9	...	6.5	3.3	9.8	0.242	6.2	2.7	8.9	5.4	0	
82	51	51	33.6	...	6.4	2.6	9.1	0.253	5.0	2.4	7.4	4.1	0	

	DPOY
259	0
294	0

```
114    0
11    0
82    0
```

[5 rows x 54 columns]

```
[133]: #dpoy 2021 prediction
nba2021 = pd.read_csv('nba2021.csv')
y_predicted = DPOYLinear.predict(nba2021)
finalldpoy = pd.concat([y_predicted,nba2021],axis=1)
finalldpoy = finalldpoy.rename(columns={0:'Predicted DPOY'})
finalldpoy = finalldpoy.sort_values(by='Predicted DPOY',ascending = False)
finalldpoy = finalldpoy.loc[finalldpoy['GS'] > 10]
finalldpoy['Player'].iloc[0]
```

[133]: 'Rudy Gobert'

```
[134]: finalldpoy.head(5)
```

```
[134]:      Predicted DPOY  Unnamed: 0      Player  Season Pos  Age  Tm  G  GS  \
177      0.003726      178    Rudy Gobert    2021  C   28  UTA  68  68
491      0.003559      492    Myles Turner    2021  C   24  IND  47  47
89       0.003025      90    Clint Capela    2021  C   26  ATL  61  61
82       0.002841      83    Jimmy Butler    2021  SF  31  MIA  51  51
116      0.002799      117  Anthony Davis    2021  PF  27  LAL  34  34

      MP  ...  OWS  DWS   WS  WS48  OBPM  DBPM  BPM  VORP  MVP  DPOY
177  30.9  ...  5.7  4.9  10.7  0.244   1.9   2.5  4.4   3.4    0    0
491  31.0  ...  1.1  2.3   3.4  0.112  -1.3   1.7  0.4   0.9    0    0
89   30.2  ...  4.7  3.1   7.8  0.204   2.7  -0.2  2.5   2.1    0    0
82   33.6  ...  6.4  2.6   9.1  0.253   5.0   2.4  7.4   4.1    0    0
116  32.2  ...  1.2  2.3   3.4  0.151   3.0   1.7  4.7   1.9    0    0
```

[5 rows x 54 columns]

```
[135]: #smoy 2021 prediction
nba2021smoy = pd.read_csv('smoy2021.csv')
y_predicted = SMOYLinear.predict(nba2021smoy)
finalismoy = pd.concat([y_predicted,nba2021smoy],axis=1)
finalismoy = finalismoy.rename(columns={0:'Predicted SMOY'})
finalismoy = finalismoy.sort_values(by='Predicted SMOY',ascending = False)
finalismoy = finalismoy.loc[finalismoy['G'] > 10]
finalismoy['Player'].iloc[0]
```

[135]: 'Jordan Clarkson'

```
[136]: finalismoy.head(5)
```

```
[136]: Predicted SMOY Unnamed: 0 Player Season Pos Age Tm G \
54 0.097336 55 Jordan Clarkson 2021 SG 28 UTA 65
139 0.084820 140 Joe Ingles 2021 SF 33 UTA 64
115 0.069711 116 Tim Hardaway Jr. 2021 SG 28 DAL 67
26 0.063696 27 Chris Boucher 2021 C 28 TOR 60
218 0.061187 219 Patty Mills 2021 PG 32 SAS 65
```

```
GS MP ... USG OWS DWS WS WS48 OBPM DBPM BPM VORP SMOY
54 1 26.6 ... 29.6 1.3 2.3 3.6 0.101 1.6 -1.0 0.6 1.1 0
139 27 27.8 ... 16.5 4.7 2.1 6.9 0.185 3.0 0.7 3.7 2.5 0
115 28 28.4 ... 23.4 2.9 1.1 4.1 0.102 1.8 -1.8 0.0 1.0 0
26 14 24.2 ... 20.5 4.0 2.1 6.1 0.203 2.8 0.1 2.9 1.8 0
218 0 24.8 ... 18.3 1.4 0.9 2.3 0.068 0.2 -1.7 -1.5 0.2 0
```

[5 rows x 53 columns]

```
[137]: #MIP 2021 prediction
nba2021mip = pd.read_csv('mip2021.csv')
y_predicted = MIPLinear.predict(nba2021mip)
finallmip = pd.concat([y_predicted,nba2021mip],axis=1)
finallmip = finallmip.rename(columns={0:'Predicted MIP'})
finallmip = finallmip.sort_values(by='Predicted MIP',ascending = False)
finallmip = finallmip.loc[finallmip['GS'] > 1]
finallmip['Player'].iloc[0]
```

```
[137]: 'Michael Porter Jr.'
```

```
[138]: finallmip.head(10)
```

```
[138]: Predicted MIP Player Season G GS MP FGM FGA \
246 0.039612 Michael Porter Jr. 2021.0 4.0 44.0 15.4 3.9 6.5
230 0.035564 Malik Beasley 2021.0 -18.0 22.0 10.8 3.0 6.5
291 0.034211 Robert Williams 2021.0 23.0 12.0 5.5 1.4 2.0
247 0.033390 Mikal Bridges 2021.0 -4.0 37.0 5.0 1.6 2.8
225 0.032606 Luguentz Dort 2021.0 16.0 24.0 6.9 2.4 6.2
214 0.030863 Kyle Kuzma 2021.0 4.0 23.0 3.8 0.2 0.3
208 0.029583 Kevin Porter Jr. 2021.0 -24.0 20.0 8.9 2.3 5.7
45 0.029120 Chris Boucher 2021.0 -2.0 14.0 11.0 2.5 4.5
153 0.025760 Jakob Poeltl 2021.0 0.0 30.0 8.9 1.3 2.1
350 0.025653 Wayne Ellington 2021.0 10.0 30.0 6.5 1.5 2.5
```

```
FGP X3PM ... TOVP USG OWS DWS WS WS48 OBPM DBPM BPM VORP
246 0.039 1.7 ... -2.5 -0.6 3.0 0.9 4.0 0.022 1.7 -0.1 1.6 1.6
230 0.015 1.6 ... -0.6 2.2 0.5 -0.3 0.2 0.004 2.4 -0.9 1.5 0.4
291 -0.006 0.0 ... -1.7 1.4 2.3 0.9 3.3 0.003 2.2 -1.2 1.0 1.3
247 0.025 0.9 ... -4.7 2.1 2.5 0.2 2.7 0.045 2.7 -0.8 2.0 1.2
225 -0.007 1.4 ... 1.6 7.6 -0.8 0.2 -0.5 -0.039 1.4 -1.4 0.0 -0.3
```


214	0.002	0.7	...	1.2	-2.4	0.2	0.9	1.1	0.015	1.3	0.5	1.8	0.8
208	-0.017	0.8	...	1.5	4.9	0.0	-0.3	-0.3	-0.011	1.9	-1.8	0.0	0.1
45	0.042	0.9	...	-0.7	0.8	1.9	0.3	2.3	-0.021	1.8	-0.9	0.9	1.0
153	-0.001	0.0	...	0.3	0.2	0.7	0.9	1.5	-0.012	-1.0	-0.8	-1.8	-0.1
350	0.090	1.1	...	-0.6	1.7	1.1	0.3	1.5	0.054	3.2	-0.8	2.5	0.5

[10 rows x 48 columns]

```
[139]: #ROY 2021 prediction
nba2021roy = pd.read_csv('roy2021.csv')
y_predicted = ROYLinear.predict(nba2021roy)
finallroy = pd.concat([y_predicted,nba2021roy],axis=1)
finallroy = finallroy.rename(columns={0:'Predicted ROY'})
finallroy = finallroy.sort_values(by='Predicted ROY',ascending = False)
finallroy = finallroy.loc[finallroy['GS'] > 1]
finallroy['Player'].iloc[0]
```

[139]: 'LaMelo Ball'

```
[140]: finallroy.head(5)
```

```
[140]:
```

	Predicted ROY	Unnamed: 0	Player	Season	Pos	Age	Tm	G	\
50	0.251742	51	LaMelo Ball	2021	PG	19	CHO	48	
35	0.214883	36	Jae'Sean Tate	2021	SF	25	HOU	67	
3	0.178309	4	Anthony Edwards	2021	SG	19	MIN	69	
83	0.132846	84	Tyrese Haliburton	2021	PG	20	SAC	58	
31	0.130995	32	Isaac Okoro	2021	SG	20	CLE	64	

	GS	MP	...	USG	OWS	DWS	WS	WS48	OBPM	DBPM	BPM	VORP	ROY
50	28	28.8	...	26.0	1.0	1.8	2.8	0.096	1.1	0.6	1.7	1.3	0
35	55	29.4	...	16.2	2.1	1.5	3.7	0.090	-1.7	0.5	-1.2	0.4	0
3	52	31.9	...	27.0	-1.0	1.4	0.4	0.008	-1.2	-1.8	-3.0	-0.5	0
83	20	30.1	...	18.1	2.8	0.6	3.5	0.096	1.9	-0.9	1.1	1.4	0
31	64	32.3	...	13.8	0.0	0.9	1.0	0.023	-4.2	-0.7	-4.9	-1.5	0

[5 rows x 53 columns]

```
[ ]:
```