```
In [1]: import pandas as pd
In [2]: df = pd.read_csv('player_mvp_stats.csv')
In [3]: df
```

Out[3]:

	Unnamed: 0	Rk	Player	Pos	Age	Team	G	GS	MP	FG	 Pts Won	Pts Max	Share	w	L	W/L%	GB	PS/G	PA/G
0	0	135.0	A.C. Green	PF	27.0	LAL	82.0	21.0	26.4	3.1	 0.0	0.0	0.0	58.0	24.0	0.707	5.0	106.3	99.6
1	1	301.0	Byron Scott	SG	29.0	LAL	82.0	82.0	32.1	6.1	 0.0	0.0	0.0	58.0	24.0	0.707	5.0	106.3	99.6
2	2	51.0	Elden Campbell	PF	22.0	LAL	52.0	0.0	7.3	1.1	 0.0	0.0	0.0	58.0	24.0	0.707	5.0	106.3	99.6
3	3	330.0	Irving Thomas	PF	25.0	LAL	26.0	0.0	4.2	0.7	 0.0	0.0	0.0	58.0	24.0	0.707	5.0	106.3	99.6
4	4	384.0	James Worthy	SF	29.0	LAL	78.0	74.0	38.6	9.2	 0.0	0.0	0.0	58.0	24.0	0.707	5.0	106.3	99.6
16389	16389	330.0	Jordan McLaughlin	PG	23.0	MIN	30.0	2.0	19.7	2.9	 0.0	0.0	0.0	19.0	45.0	0.297	22.5	113.3	117.5
16390	16390	381.0	Josh Okogie	SG	21.0	MIN	62.0	28.0	25.0	2.7	 0.0	0.0	0.0	19.0	45.0	0.297	22.5	113.3	117.5
16391	16391	477.0	Karl- Anthony Towns	С	24.0	MIN	35.0	35.0	33.9	9.0	 0.0	0.0	0.0	19.0	45.0	0.297	22.5	113.3	117.5
16392	16392	317.0	Kelan Martin	SF	24.0	MIN	31.0	4.0	16.0	2.3	 0.0	0.0	0.0	19.0	45.0	0.297	22.5	113.3	117.5
16393	16393	420.0	Naz Reid	С	20.0	MIN	30.0	11.0	16.5	3.3	 0.0	0.0	0.0	19.0	45.0	0.297	22.5	113.3	117.5

16394 rows × 42 columns

```
In [4]: # to do data cleaning before ml
del df['Unnamed: 0']
```

```
In [232]: df.isna().sum().head(2)
Out[232]: Rk
          Player
                     0
          dtype: int64
In [18]: |df[df['3P%'].isna()][['3PA','3P']].value counts()
Out[18]: 3PA 3P
          0.0 0.0
                       2397
          dtype: int64
            • all players with missing 3P% have zero attemps
In [19]: | df = df.fillna(0)
In [233]: df.isna().sum().head(2)
Out[233]: Rk
                     0
          Player
                     0
          dtype: int64
In [23]: df.columns
Out[23]: Index(['Rk', 'Player', 'Pos', 'Age', 'Team', 'G', 'GS', 'MP', 'FG', 'FGA',
                  'FG%', '3P', '3PA', '3P%', '2P', '2PA', '2P%', 'eFG%', 'FT', 'FTA',
                  'FT%', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK', 'TOV', 'PF', 'PTS',
                  'Year', 'Pts Won', 'Pts Max', 'Share', 'W', 'L', 'W/L%', 'GB', 'PS/G',
                  'PA/G', 'SRS'],
                dtype='object')
In [24]: # get ride of columns that directly affect the y, causing overfit problem
          predictors = ['Age','G', 'GS', 'MP', 'FG', 'FGA',
                  'FG%', '3P', '3PA', '3P%', '2P', '2PA', '2P%', 'eFG%', 'FT', 'FTA',
                 'FT%', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK', 'TOV', 'PF', 'PTS',
                  'Year','W', 'L', 'W/L%', 'GB', 'PS/G',
                  'PA/G', 'SRS']
```

```
In [25]: train = df.query('Year < 2021')</pre>
In [27]: test = df.query('Year == 2021')
In [29]: from sklearn.linear_model import Ridge # ridge prevent overfit
In [31]: reg = Ridge(alpha=0.1)
In [47]: reg.fit(train[predictors],train['Share'])
Out[47]:
                Ridge
          Ridge(alpha=0.1)
In [50]: test.shape
Out[50]: (619, 41)
In [51]: prediction_1s = reg.predict(test[predictors])
In [52]: pd_pred = pd.DataFrame(prediction_1s,columns=['prediction'],index = test.index)
```

```
In [53]: pd_pred
```

Out[53]:

	prediction
1175	0.014023
1176	0.009959
1177	-0.013762
1178	-0.017826
1179	0.007912
16117	-0.008192
16118	-0.000126
16119	-0.003047
16120	-0.023829
16121	-0.006096
610 rov	vo v 1 oolum

619 rows × 1 columns

```
In [68]: combination = pd.concat([pd_pred,test[['Share','Player']]],axis=1)
```

In [69]: combination

Out[69]:

	prediction	Share	Player
1175	0.014023	0.0	Aaron Gordon
1176	0.009959	0.0	Aaron Gordon
1177	-0.013762	0.0	Al-Farouq Aminu
1178	-0.017826	0.0	Al-Farouq Aminu
1179	0.007912	0.0	Alex Len
16117	-0.008192	0.0	Rodney McGruder
16118	-0.000126	0.0	Saben Lee
16119	-0.003047	0.0	Saddiq Bey
16120	-0.023829	0.0	Sekou Doumbouya
16121	-0.006096	0.0	Wayne Ellington

619 rows × 3 columns

```
In [75]: combination.sort_values('Share', ascending=False).head(10)
```

Out[75]:

	prediction	Share	Player
13532	0.145136	0.961	Nikola Jokić
10473	0.156836	0.580	Joel Embiid
5089	0.139247	0.449	Stephen Curry
12018	0.200518	0.345	Giannis Antetokounmpo
2450	0.068735	0.138	Chris Paul
13156	0.145498	0.042	Luka Dončić
9208	0.110186	0.038	Damian Lillard
4983	0.083648	0.020	Julius Randle
1223	0.028737	0.010	Derrick Rose
1224	0.024673	0.010	Derrick Rose

• how to improve model: identify an error metric

In [80]: combination

Out[80]:

	prediction	Share	Player	Rk
13532	0.145136	0.961	Nikola Jokić	1
10473	0.156836	0.580	Joel Embiid	2
5089	0.139247	0.449	Stephen Curry	3
12018	0.200518	0.345	Giannis Antetokounmpo	4
2450	0.068735	0.138	Chris Paul	5
2720	-0.000281	0.000	Devon Dotson	615
2721	-0.031390	0.000	Garrett Temple	616
2722	0.018648	0.000	Lauri Markkanen	617
2723	-0.019941	0.000	Patrick Williams	618
16121	-0.006096	0.000	Wayne Ellington	619

619 rows × 4 columns

```
In [82]: combination = combination.sort_values('prediction',ascending=False)
In [83]: combination['Predict_Rk'] = list(range(1,combination.shape[0]+1))
```

In [84]: combination

Out[84]:

	prediction	Share	Player	Rk	Predict_Rk
12018	0.200518	0.345	Giannis Antetokounmpo	4	1
10473	0.156836	0.580	Joel Embiid	2	2
13156	0.145498	0.042	Luka Dončić	6	3
13532	0.145136	0.961	Nikola Jokić	1	4
5162	0.140184	0.001	LeBron James	17	5
1296	-0.047716	0.000	P.J. Tucker	376	615
16115	-0.048117	0.000	Killian Hayes	212	616
11457	-0.048408	0.000	Patrick McCaw	77	617
9669	-0.052177	0.000	Anžejs Pasečņiks	126	618
13928	-0.053230	0.000	Didi Louzada	186	619

619 rows × 5 columns

```
In [86]: combination.sort_values('Share', ascending=False).head(10)
```

Out[86]:

	prediction	Share	Player	Rk	Predict_Rk
13532	0.145136	0.961	Nikola Jokić	1	4
10473	0.156836	0.580	Joel Embiid	2	2
5089	0.139247	0.449	Stephen Curry	3	6
12018	0.200518	0.345	Giannis Antetokounmpo	4	1
2450	0.068735	0.138	Chris Paul	5	36
13156	0.145498	0.042	Luka Dončić	6	3
9208	0.110186	0.038	Damian Lillard	7	13
4983	0.083648	0.020	Julius Randle	8	28
1224	0.024673	0.010	Derrick Rose	10	108
1223	0.028737	0.010	Derrick Rose	9	92

• to see how long to take to include that person

```
In [96]: find_ap(combination)
Out[96]: 0.7111111111111
In [94]: x
```

Out[94]:

	prediction	Share	Player	Rk	Predict_Rk
12018	0.200518	0.345	Giannis Antetokounmpo	4	1
10473	0.156836	0.580	Joel Embiid	2	2
13156	0.145498	0.042	Luka Dončić	6	3
13532	0.145136	0.961	Nikola Jokić	1	4
5162	0.140184	0.001	LeBron James	17	5
1296	-0.047716	0.000	P.J. Tucker	376	615
16115	-0.048117	0.000	Killian Hayes	212	616
11457	-0.048408	0.000	Patrick McCaw	77	617
9669	-0.052177	0.000	Anžejs Pasečņiks	126	618
13928	-0.053230	0.000	Didi Louzada	186	619

619 rows × 5 columns

```
In [97]: years = list(range(1991,2022))
```

```
In [111]: aps = []
          all prediction = []
          for year in years[5:]:
              train = df[df['Year'] < year]</pre>
              test = df[df['Year'] == year]
              reg.fit(train[predictors],train['Share'])
              predictions = reg.predict(test[predictors])
              predictions = pd.DataFrame(predictions,columns=['prediction'],index = test.index)
              combination = pd.concat([test[['Share','Player']],predictions],axis=1)
              all prediction.append(combination)
              aps.append(find ap(combination))
In [123]: all prediction
                                   Player prediction
                   Share
           513
                     0.0
                              A.J. Guyton
                                             -0.007910
           514
                     0.0
                           Charles Oakley
                                             -0.024502
           515
                          Dalibor Bagarić
                     0.0
                                             -0.004729
           516
                           Eddie Robinson
                                             0.000664
                     0.0
           517
                     0.0
                               Eddy Curry
                                             -0.000428
           . . .
                     . . .
                                                   . . .
           15645
                     0.0
                            Steve Francis
                                              0.028624
                           Terence Morris
                                             -0.015437
           15646
                     0.0
           15647
                             Tierre Brown
                     0.0
                                             -0.003299
           15648
                            Walt Williams
                     0.0
                                             -0.013648
           15649
                             Óscar Torres
                     0.0
                                             -0.007015
           [500 rows x 3 columns],
                                   Player prediction
                   Share
                              A.J. Guyton
           523
                     0.0
                                              0.007825
           524
                             Adonal Foyle
                     0.0
                                              0.003776
           525
                     0.0
                           Antawn Jamison
                                              0.037919
           526
                                 Bob Sura
                     0.0
                                              0.002127
           527
                     0.0
                                             -0.003989
                              Chris Mills
In [134]: def add ranks(combination):
              combination = combination.sort values("Share", ascending=False)
              combination['Rk'] = list(range(1,combination.shape[0]+1))
              combination = combination.sort values('prediction',ascending=False)
              combination['Predict Rk'] = list(range(1, combination.shape[0]+1))
              combination['Diff'] = combination['Rk'] - combination['Predict Rk']
              return combination
```

```
In [150]: add ranks(all prediction[1]).query('Rk < 6').sort values('Diff',ascending=False)
Out[150]:
                              Player prediction Rk Predict_Rk
                                                           Diff
                  Share
                                     0.176051
                                                        2
                                                            -1
            2642
                  0.857
                          Karl Malone
                  0.832 Michael Jordan
                                     0.154297
                                              2
                                                            -1
            12663
                  0.327
                            Grant Hill
                                     0.125242
                                                            -1
            16240
                  0.207
                        Tim Hardaway
                                     0.050239
                                                       26
                                                           -22
            5989
                            Glen Rice
                                                      107 -102
                  0.117
                                     0.019718
            10158
In [193]: def backtest(stats, model, year, predictors):
               aps = []
               all prediction = []
               for year in years[5:]:
                   train = df[df['Year'] < year]</pre>
                   test = df[df['Year'] == year]
                   model.fit(train[predictors],train['Share'])
                   predictions = reg.predict(test[predictors])
                   predictions = pd.DataFrame(predictions,columns=['prediction'],index = test.index)
                   combination = pd.concat([test[['Share','Player']],predictions],axis=1)
                   combination = add_ranks(combination)
                   all_prediction.append(combination)
                    aps.append(find_ap(combination))
               return sum(aps)/len(aps),aps,pd.concat(all prediction)
In [234]: mean ap, aps, all prediction = backtest(df, reg, years[5:], predictors)
In [191]: mean ap
Out[191]: 0.6944979748368247
```

· Diagnosing model performance

In [154]: all_prediction[all_prediction['Rk']<=5].sort_values('Diff').head(10)</pre>

Out[154]:

	Share	Player	prediction	Rk	Predict_Rk	Diff
10158	0.117	Glen Rice	0.019718	5	107	-102
2055	0.712	Jason Kidd	0.023461	2	82	-80
6672	0.839	Steve Nash	0.036889	1	49	-48
5095	0.344	Chauncey Billups	0.044579	5	41	-36
14815	0.258	Joakim Noah	0.043823	4	39	-35
2450	0.138	Chris Paul	0.068735	5	36	-31
6687	0.739	Steve Nash	0.056451	1	30	-29
10379	0.228	Peja Stojaković	0.039963	4	29	-25
5989	0.207	Tim Hardaway	0.050239	4	26	-22
6701	0.785	Steve Nash	0.071394	2	21	-19

In [236]: pd.concat([pd.Series(reg.coef_),pd.Series(predictors)],axis=1).sort_values(0,ascending=False).head(2)

Out[236]:

34 0.051508 PTS_R

· adding more predictors

```
In [165]: | df.columns
Out[165]: Index(['Rk', 'Player', 'Pos', 'Age', 'Team', 'G', 'GS', 'MP', 'FG', 'FGA',
                   'FG%', '3P', '3PA', '3P%', '2P', '2PA', '2P%', 'eFG%', 'FT', 'FTA',
                   'FT%', 'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK', 'TOV', 'PF', 'PTS',
                   'Year', 'Pts Won', 'Pts Max', 'Share', 'W', 'L', 'W/L%', 'GB', 'PS/G',
                   'PA/G', 'SRS'],
                  dtype='object')
In [179]: df ratios = df[['PTS','AST','STL','BLK','3P','Year']].groupby('Year').apply(lambda x:x/x.mean())
In [186]: df[['PTS R','AST R','STL R','BLK R','3P R']] = df ratios[['PTS','AST','STL','BLK','3P']]
In [188]: df.head()
Out[188]:
                Rk
                      Player Pos Age Team
                                                GS
                                                     MP FG FGA ... W/L% GB PS/G PA/G SRS
                                                                                                 PTS R
                                                                                                         AST R
                                                                                                                          BLK
                                             G
                                                                                                                  STL R
                        A.C.
                             PF 27.0
            0 135.0
                                       LAL 82.0 21.0 26.4 3.1
                                                             6.6 ... 0.707 5.0 106.3
                                                                                     99.6 6.73 1.019207 0.451512 0.984209 0.7029
                       Green
                       Byron
                             SG 29.0
            1 301.0
                                       LAL 82.0 82.0 32.1 6.1 12.8 ... 0.707 5.0 106.3
                                                                                     99.6 6.73 1.624011 1.103695 1.687215 0.7029
                       Scott
                       Elden
                             PF 22.0
                                       LAL 52.0
                                                0.0 7.3 1.1 2.4 ... 0.707 5.0 106.3 99.6 6.73 0.313602 0.100336 0.281203 1.640<sup>-1</sup>
                    Campbell
                       Irving
            3 330.0
                             PF 25.0
                                       LAL 26.0
                                                0.0 4.2 0.7
                                                             1.9 ... 0.707 5.0 106.3
                                                                                     99.6 6.73 0.201601 0.200672 0.281203 0.0000
                     Thomas
                      James
            4 384.0
                             SF 29.0
                                       LAL 78.0 74.0 38.6 9.2 18.7 ... 0.707 5.0 106.3 99.6 6.73 2.396817 1.755878 1.827817 0.9372
                      Worthy
           5 rows × 46 columns
In [195]: predictors += ['PTS R', 'AST R', 'STL_R', 'BLK_R', '3P_R']
In [208]: del predictors[-6]
```

```
In [238]: predictors
Out[238]: ['Age',
            'G',
             'GS',
             'MP',
             'FG',
            'FGA',
            'FG%',
             '3P',
            '3PA',
            '3P%',
            '2P',
            '2PA',
            '2P%',
            'eFG%',
             'FT',
            'FTA',
            'FT%',
            'ORB',
            'DRB',
            'TRB',
             'AST',
             'STL',
            'BLK',
            'TOV',
            'PF',
            'PTS',
            'Year',
             'W',
             'L',
             'W/L%',
             'GB',
             'PS/G',
            'PA/G',
             'SRS',
             'PTS_R',
             'AST_R',
             'STL R',
             'BLK R',
             '3P_R']
```

```
In [210]: mean ap,aps,all prediction = backtest(df,reg,years[5:],predictors)
In [211]: mean_ap
Out[211]: 0.6933074406516423
In [218]: df['NPos'] = df['Pos'].astype('category').cat.codes
In [220]: df['NTm'] = df['Team'].astype('category').cat.codes

    using random forest

In [230]: from sklearn.ensemble import RandomForestRegressor
          rf = RandomForestRegressor(n_estimators=50,random_state=1,min_samples_split=5)
          mean_ap1,aps,all_prediction = backtest(df,rf,years[28:],predictors)
In [231]: mean_ap1
Out[231]: 0.7392204084879423
In [228]: mean_ap,aps,all_prediction = backtest(df,reg,years[28:],predictors)
In [229]: mean_ap
Out[229]: 0.6933074406516423
  In [ ]:
```