6.5.2 Forward and Backward Stepwise Selection

May 18, 2018

In [44]:
```
# conventional way to import pandas
import pandas as pd
# conventional way to import seaborn
import seaborn as sns
# conventional way to import numpy
import numpy as np

from sklearn import metrics
import matplotlib.pyplot as plt

data = pd.read_csv("https://vincentarelbundock.github.io/Rdatasets/csv/ISLR/Hitters.cs

data.head()
```

Out[44]:

	AtBat	Hits	HmRun	Runs	RBI	Walks	Years	CAtBat	CHits	\
-Andy Allanson	293	66	1	30	29	14	1	293	66	
-Alan Ashby	315	81	7	24	38	39	14	3449	835	
-Alvin Davis	479	130	18	66	72	76	3	1624	457	
-Andre Dawson	496	141	20	65	78	37	11	5628	1575	
-Andres Galarraga	321	87	10	39	42	30	2	396	101	

	CHmRun	CRuns	CRBI	CWalks	League	Division	PutOuts	\
-Andy Allanson	1	30	29	14	A	E	446	
-Alan Ashby	69	321	414	375	N	W	632	
-Alvin Davis	63	224	266	263	A	W	880	
-Andre Dawson	225	828	838	354	N	E	200	
-Andres Galarraga	12	48	46	33	N	E	805	

	Assists	Errors	Salary	NewLeague
-Andy Allanson	33	20	NaN	A
-Alan Ashby	43	10	475.0	N
-Alvin Davis	82	14	480.0	A
-Andre Dawson	11	3	500.0	N
-Andres Galarraga	40	4	91.5	N

After listing the data we can see that some have missing data for their Salary. Next drop all the rows that contain NaN data.

```
In [45]: data = data.dropna()
         data.index.name = 'Player'
         data.head()
```

Out[45]:

Player	AtBat	Hits	HmRun	Runs	RBI	Walks	Years	CAtBat	CHits
-Alan Ashby	315	81	7	24	38	39	14	3449	835
-Alvin Davis	479	130	18	66	72	76	3	1624	457
-Andre Dawson	496	141	20	65	78	37	11	5628	1575
-Andres Galarraga	321	87	10	39	42	30	2	396	101
-Alfredo Griffin	594	169	4	74	51	35	11	4408	1133

Player	CHmRun	CRuns	CRBI	CWalks	League	Division	PutOuts
-Alan Ashby	69	321	414	375	N	W	632
-Alvin Davis	63	224	266	263	A	W	880
-Andre Dawson	225	828	838	354	N	E	200
-Andres Galarraga	12	48	46	33	N	E	805
-Alfredo Griffin	19	501	336	194	A	W	282

Player	Assists	Errors	Salary	NewLeague
-Alan Ashby	43	10	475.0	N
-Alvin Davis	82	14	480.0	A
-Andre Dawson	11	3	500.0	N
-Andres Galarraga	40	4	91.5	N
-Alfredo Griffin	421	25	750.0	A

We will have to do this by hand using the following code and display infomation about the variables we converted from strings to numbers.

```
In [46]: dummieVariables = pd.get_dummies(data[['League', 'Division', 'NewLeague']])
         dummieVariables.info()
         print(dummieVariables.head())
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 263 entries, -Alan Ashby to -Willie Wilson
Data columns (total 6 columns):
League_A       263 non-null uint8
League_N       263 non-null uint8
Division_E     263 non-null uint8
Division_W     263 non-null uint8
NewLeague_A    263 non-null uint8
NewLeague_N    263 non-null uint8
dtypes: uint8(6)
memory usage: 3.6+ KB
                  League_A  League_N  Division_E  Division_W  NewLeague_A  \
Player
-Alan Ashby               0         1           0           1            0
```

```
-Alvin Davis              1          0           0           1            1
-Andre Dawson             0          1           1           0            0
-Andres Galarraga         0          1           1           0            0
-Alfredo Griffin          1          0           0           1            1

                   NewLeague_N
Player
-Alan Ashby                  1
-Alvin Davis                 0
-Andre Dawson                1
-Andres Galarraga            1
-Alfredo Griffin             0
```

Next we must remove the columns with our independent variable (Salary), and columns for which we created dummy variables and reinterduce them into our predictors.

```
In [47]: dummieVariables = pd.get_dummies(data[['League', 'Division', 'NewLeague']])
         dummieVariables.info()
         print(dummieVariables.head())

         y = data.Salary
         X_ = data.drop(['Salary', 'League', 'Division', 'NewLeague'], axis=1).astype('float64

         X = pd.concat([X_, dummieVariables[['League_N', 'Division_W', 'NewLeague_N']]], axis=
         X.head()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 263 entries, -Alan Ashby to -Willie Wilson
Data columns (total 6 columns):
League_A       263 non-null uint8
League_N       263 non-null uint8
Division_E     263 non-null uint8
Division_W     263 non-null uint8
NewLeague_A    263 non-null uint8
NewLeague_N    263 non-null uint8
dtypes: uint8(6)
memory usage: 3.6+ KB
                   League_A  League_N  Division_E  Division_W  NewLeague_A  \
Player
-Alan Ashby               0         1           0           1            0
-Alvin Davis              1         0           0           1            1
-Andre Dawson             0         1           1           0            0
-Andres Galarraga         0         1           1           0            0
-Alfredo Griffin          1         0           0           1            1

                   NewLeague_N
Player
```

```
-Alan Ashby                 1
-Alvin Davis                0
-Andre Dawson               1
-Andres Galarraga           1
-Alfredo Griffin            0
```

Out[47]:

Player	AtBat	Hits	HmRun	Runs	RBI	Walks	Years	CAtBat
-Alan Ashby	315.0	81.0	7.0	24.0	38.0	39.0	14.0	3449.0
-Alvin Davis	479.0	130.0	18.0	66.0	72.0	76.0	3.0	1624.0
-Andre Dawson	496.0	141.0	20.0	65.0	78.0	37.0	11.0	5628.0
-Andres Galarraga	321.0	87.0	10.0	39.0	42.0	30.0	2.0	396.0
-Alfredo Griffin	594.0	169.0	4.0	74.0	51.0	35.0	11.0	4408.0

Player	CHits	CHmRun	CRuns	CRBI	CWalks	PutOuts	Assists
-Alan Ashby	835.0	69.0	321.0	414.0	375.0	632.0	43.0
-Alvin Davis	457.0	63.0	224.0	266.0	263.0	880.0	82.0
-Andre Dawson	1575.0	225.0	828.0	838.0	354.0	200.0	11.0
-Andres Galarraga	101.0	12.0	48.0	46.0	33.0	805.0	40.0
-Alfredo Griffin	1133.0	19.0	501.0	336.0	194.0	282.0	421.0

Player	Errors	League_N	Division_W	NewLeague_N
-Alan Ashby	10.0	1	1	1
-Alvin Davis	14.0	0	1	0
-Andre Dawson	3.0	1	0	1
-Andres Galarraga	4.0	1	0	1
-Alfredo Griffin	25.0	0	1	0

bla bla

In [48]:
```
import statsmodels.api as sm
import itertools
max_predictors = 3

# Functions found at "https://github.com/qx0731/ISL_python/blob/master/Chapter_6_sec_
def CalulateRSS(y, X, predictors_list):
    model = sm.OLS(y, X[list(predictors_list)]).fit()
    RSS   = ((model.predict(X[list(predictors_list)]) - y) ** 2)
    RSS   = RSS.sum()
    return {'Model':model, "RSS":RSS}

def forwardStepwiseSelection(y, X, predictors_list):
    remaining_predictors = [p for p in X.columns if p not in predictors_list]
    results = []
    for p in remaining_predictors:
```

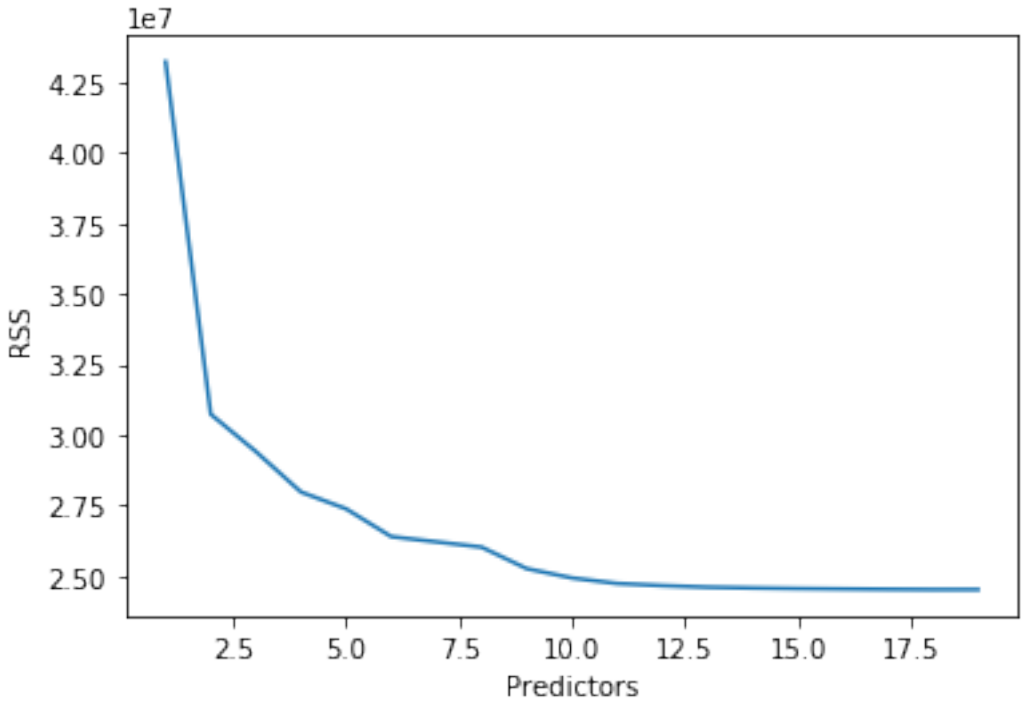
```
            results.append(CalulateRSS(y, X, feature_list+[p]))

        models = pd.DataFrame(results)
        best_model_forwardsetwise = models.loc[models['RSS'].argmin()]
        return best_model_forwardsetwise

    def backwardStepwiseSelection(y, X, predictors_list):
        results = []
        for combo in itertools.combinations(predictors_list, len(predictors_list)-1):
            results.append(CalulateRSS(y, X, combo))

        models = pd.DataFrame(results)
        best_model = models.loc[models['RSS'].argmin()]
        return best_model
```

bla bla

```
In [49]: forwardModels = pd.DataFrame(columns=["RSS", "Model"])
         feature_list = []
         for i in range(1,len(X.columns)+1):
             forwardModels.loc[i] = forwardStepwiseSelection(y, X, feature_list)
             feature_list = forwardModels.loc[i]["Model"].model.exog_names
```

```
C:\Users\au479931\AppData\Local\Continuum\anaconda3\lib\site-packages\ipykernel_launcher.py:19
```

```
In [50]: plt.figure()
         plt.plot(forwardModels["RSS"])
         plt.xlabel('Predictors')
         plt.ylabel('RSS')
         plt.show()
```




```
In [51]: print(forwardModels.loc[max_predictors, 'Model'].params)
```

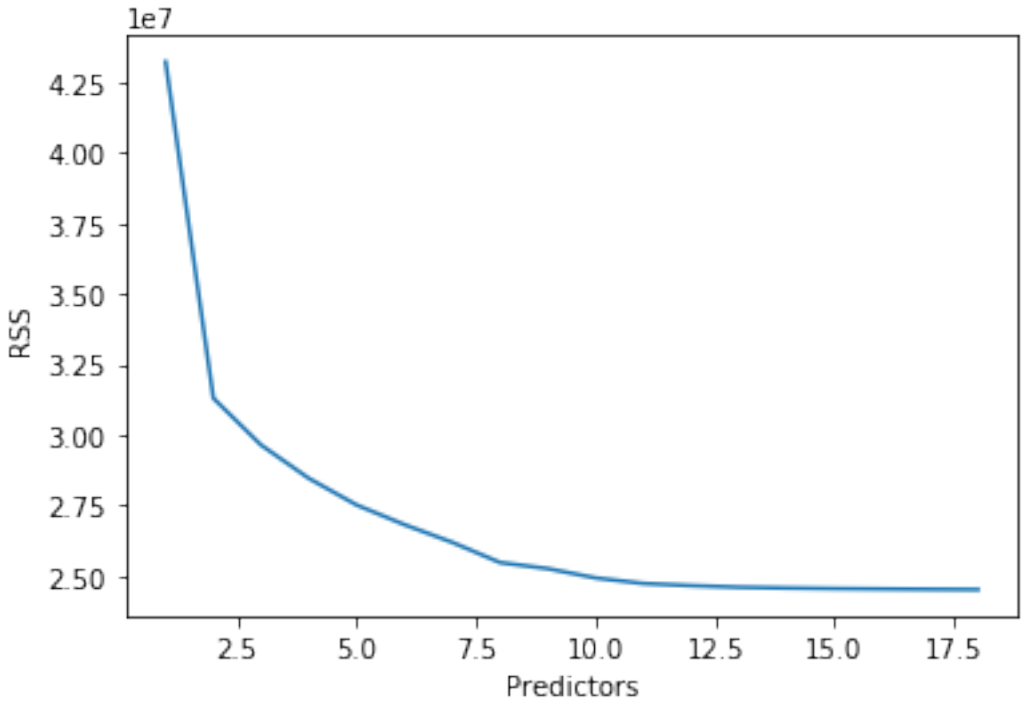
```
Hits            3.405706
CRBI            0.696362
Division_W   -129.160367
dtype: float64
```

bla bla

```
In [52]: backwardModels = pd.DataFrame(columns=["RSS", "Model"], index = range(1,len(X.columns)
         feature_list = X.columns

         while(len(feature_list) > 1):
             backwardModels.loc[len(feature_list)-1] = backwardStepwiseSelection(y, X, feature_
             feature_list = backwardModels.loc[len(feature_list)-1]["Model"].model.exog_names
```

```
C:\Users\au479931\AppData\Local\Continuum\anaconda3\lib\site-packages\ipykernel_launcher.py:28
```

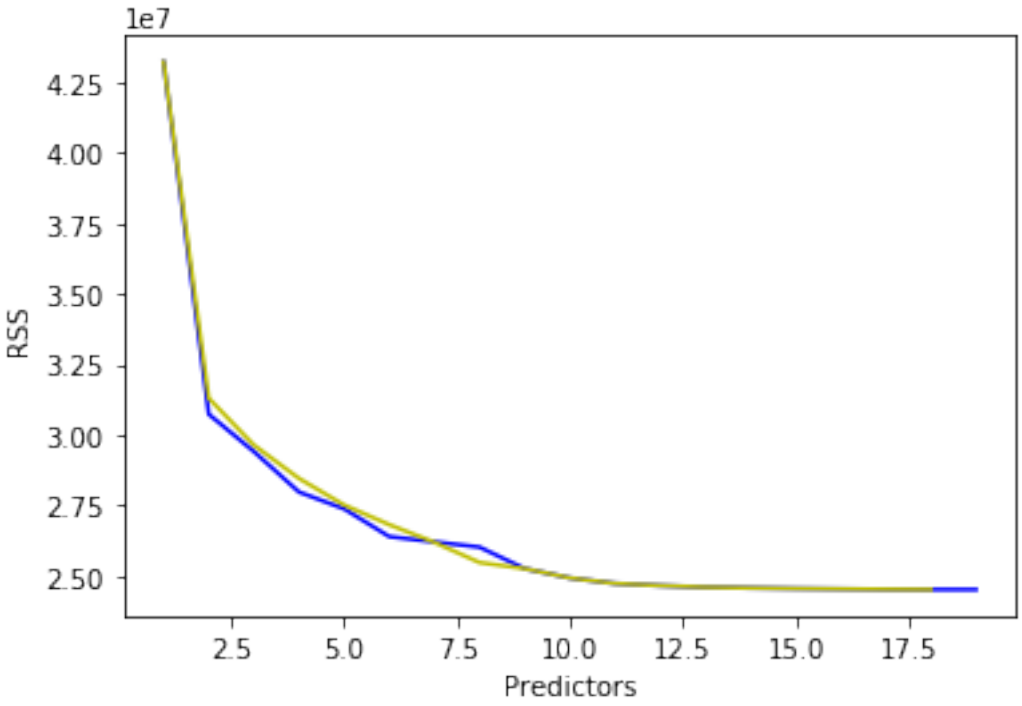
```
In [53]: plt.figure()
         plt.plot(backwardModels["RSS"])
         plt.xlabel('Predictors')
         plt.ylabel('RSS')
         plt.show()
```




```
In [54]: print(backwardModels.loc[max_predictors, "Model"].params)

Hits       2.111712
CRuns      0.646149
PutOuts    0.295625
dtype: float64
```

Plotting both Forward and Backward Stepwise Selection in a single plot, to see the difference. Forward is blue, Backward is yellow.

```
In [56]: plt.figure()
         plt.plot(forwardModels["RSS"],'b')
         plt.plot(backwardModels["RSS"],'y')
         plt.xlabel('Predictors')
         plt.ylabel('RSS')
         plt.show()
```


1e7
4.25
4.00
3.75
3.50
3.25
3.00
2.75
2.50
RSS
2.5
5.0
7.5
10.0
12.5
15.0
17.5
Predictors