12/5/21, 11:58 PM HW5 Problem6

```
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
         from sklearn.metrics import confusion matrix
          from sklearn.model selection import train test split
          from sklearn.model selection import LeaveOneOut
          from sklearn.linear model import LinearRegression
          from sklearn.neighbors import KNeighborsRegressor
          from sklearn.ensemble import RandomForestRegressor
          from sklearn.metrics import mean squared error
          from sklearn.tree import DecisionTreeRegressor
          from sklearn.model selection import KFold
          from sklearn.ensemble import BaggingRegressor
          from sklearn.ensemble import GradientBoostingRegressor
In [2]:
         df = pd.read csv('Hitters.csv')
          df copy = df.copy()
          df copy['League'] = df copy['League'].replace({'N': 1.0, 'A': 0.0})
          df copy['Division'] = df copy['Division'].replace({'W': 1.0, 'E': 0.0})
          df copy['NewLeague'] = df copy['NewLeague'].replace({'N': 1.0, 'A': 0.0})
          df copy.head()
Out[2]:
            AtBat Hits HmRun Runs RBI Walks Years CAtBat CHits CHmRun CRuns CRBI CWalks League Division PutOuts Assists Errors
         0
             293
                   66
                            1
                                 30
                                      29
                                             14
                                                    1
                                                          293
                                                                 66
                                                                                 30
                                                                                       29
                                                                                               14
                                                                                                      0.0
                                                                                                               0.0
                                                                                                                       446
                                                                                                                               33
                                                                                                                                      20
             315
                   81
                            7
                                 24
                                      38
                                             39
                                                   14
                                                         3449
                                                                835
                                                                          69
                                                                                 321
                                                                                      414
                                                                                              375
                                                                                                      1.0
                                                                                                               1.0
                                                                                                                       632
                                                                                                                               43
                                                                                                                                      10
             479
                  130
                                 66
                                      72
                                             76
                                                         1624
                                                                457
                                                                          63
                                                                                 224
                                                                                      266
                                                                                              263
                                                                                                      0.0
                                                                                                               1.0
                                                                                                                       880
                                                                                                                               82
                                                                                                                                      14
                            18
         3
             496
                            20
                                 65
                                      78
                                             37
                                                   11
                                                         5628
                                                               1575
                                                                         225
                                                                                828
                                                                                      838
                                                                                              354
                                                                                                      1.0
                                                                                                               0.0
                                                                                                                       200
                                                                                                                               11
                                                                                                                                       3
                  141
             321
                   87
                            10
                                 39
                                      42
                                             30
                                                    2
                                                          396
                                                                101
                                                                          12
                                                                                 48
                                                                                       46
                                                                                               33
                                                                                                      1.0
                                                                                                               0.0
                                                                                                                       805
                                                                                                                               40
```

#### Problem a

```
In [3]: # Remove instances with missing values
df_copy = df_copy.dropna(axis=0)
```

```
# log-transform the Salary Column
df_copy['Salary'] = np.log(df_copy['Salary'])
```

### Problem b

```
In [5]: y = df_copy['Salary']

X = df_copy.drop('Salary',1)

X_train = X.iloc[:200,:]
    X_test = X.iloc[200:,:]
    y_train = y.iloc[:200]
    y_test = y.iloc[200:]
```

C:\Users\ADAMYA~1\AppData\Local\Temp/ipykernel\_28264/1672455205.py:3: FutureWarning: In a future version of pandas all ar guments of DataFrame.drop except for the argument 'labels' will be keyword-only X = df copy.drop('Salary',1)

### Problem c

```
In [6]:
         lambda list = np.linspace(0.1,0.5,50)
         training error list = []
         testing error list = []
         for lambda value in (lambda list):
             # Create Gradient Boosting model and fit to training data
             GB regressor = GradientBoostingRegressor(learning rate=lambda value, n estimators=1000)
             GB regressor.fit(X train,y train)
             training y pred = GB regressor.predict(X train)
             training error = mean squared error(y train, training y pred)
             testing y pred = GB regressor.predict(X test)
             testing error = mean squared error(y test, testing y pred)
             training error list.append(training error)
             testing error list.append(testing error)
         plt.plot(lambda list,training error list)
         plt.xlabel("Shrinkage Parameter")
         plt.ylabel("Training MSE")
```

```
Out[6]: Text(0, 0.5, 'Training MSE')

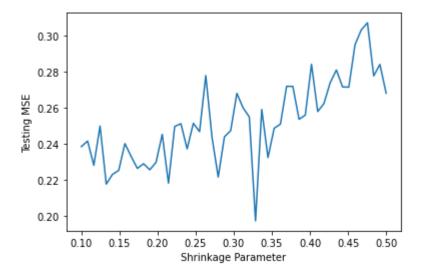
1e-9

5 - 4

4 - 1

0 - 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.45 0.50 Shrinkage Parameter
```

## Problem d



### Problem e

```
In [66]:
          # Simple Linear Regression
          LinRegr = LinearRegression()
          LinRegr.fit(X train,y train)
          y pred = LinRegr.predict(X test)
          MSE = mean_squared_error(y_test,y_pred)
          print("The MSE for Linear Regression is " + repr(MSE))
          # K-Nearest Neighbors Regression
          KN regressor = KNeighborsRegressor()
          KN regressor.fit(X train,y train)
          y pred = KN regressor.predict(X test)
          MSE = mean_squared_error(y_test,y_pred)
          print("The MSE for K-Neighbors Regression is " + repr(MSE))
          # Gradient Boosting Regression
          GB regressor = GradientBoostingRegressor()
          GB regressor.fit(X train,y train)
          y pred = GB regressor.predict(X test)
          MSE = mean squared error(y test,y pred)
          print("The MSE for Gradient Boosting Regression is " + repr(MSE))
```

```
The MSE for Linear Regression is 0.49541867535174294
The MSE for K-Neighbors Regression is 0.4518306295380096
The MSE for Gradient Boosting Regression is 0.3859412214403504
```

### Problem f

Out[8]:		Features	Feature Importances
out[o].		reatures	reature importances
	0	AtBat	0.015306
	1	Hits	0.046453
	2	HmRun	0.005153
	3	Runs	0.001224
	4	RBI	0.004384
	5	Walks	0.063386
	6	Years	0.020429
	7	CAtBat	0.587890
	8	CHits	0.031900
	9	CHmRun	0.031011
	10	CRuns	0.056883
	11	CRBI	0.014145
	12	CWalks	0.081922
	13	League	0.000102
	14	Division	0.000597

	Features	Feature Importances
15	PutOuts	0.019651
16	Assists	0.013429
17	Errors	0.004197
18	NewLeague	0.001938

# Problem g