KFold

August 1, 2019

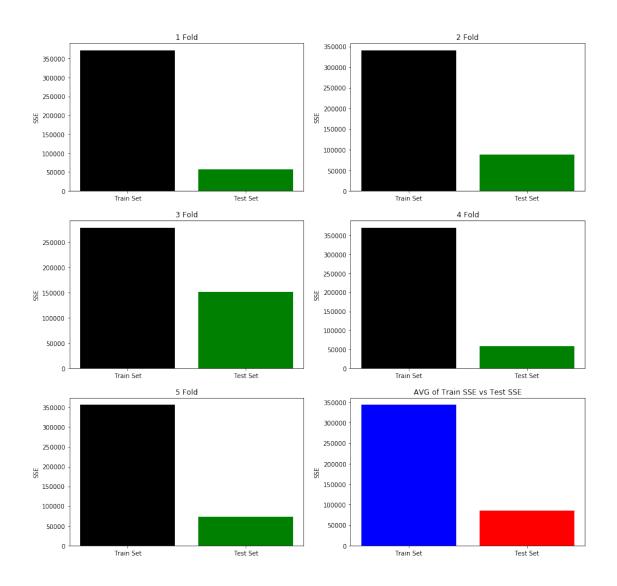
```
In [1]: import warnings
        warnings.filterwarnings("ignore")
        from matplotlib import pyplot as plt
        #from sklearn import datasets, linear_model
        from sklearn.linear_model import LinearRegression
        import pandas as pd
        from sklearn.model_selection import KFold, train_test_split
        from sklearn.preprocessing import MinMaxScaler
        from sklearn.svm import SVR
        import numpy as np
        from sklearn.cross_validation import cross_val_score, cross_val_predict
        from sklearn import metrics
/opt/conda/lib/python3.6/site-packages/sklearn/cross_validation.py:41: DeprecationWarning: This
  "This module will be removed in 0.20.", DeprecationWarning)
In [2]: def change_month(month):
            month = month.lower()
            month = month.replace('jan','1')
            month = month.replace('feb','2')
            month = month.replace('mar','3')
            month = month.replace('apr','4')
            month = month.replace('may','5')
            month = month.replace('jun','6')
            month = month.replace('jul','7')
            month = month.replace('aug','8')
            month = month.replace('sep','9')
            month = month.replace('oct','10')
            month = month.replace('nov','11')
            month = month.replace('dec','12')
            return month
        def change_day(day):
            day = day.lower()
```

```
day = day.replace('mon','1')
            day = day.replace('tue','2')
            day = day.replace('wed','3')
            day = day.replace('thu','4')
            day = day.replace('fri','5')
            day = day.replace('sat','6')
            day = day.replace('sun','7')
            return day
In [3]: df = pd.read_csv('forestfires.csv')
        # Get names of indexes for which column Age has value 30
        indexNames = df[ df['area'] <= 0].index</pre>
        # Delete these row indexes from dataFrame
        df.drop(indexNames , inplace=True)
        indexNames = df[df['area'] >400].index
        df.drop(indexNames,inplace=True)
        df['month'] = df['month'].apply(change_month)
        df['day'] = df['day'].apply(change_day)
        Y = df['area'].values
        X1 = df['FFMC'].values
        X2 = df['DMC'].values
        X3 = df['DC'].values
        X4 = df['ISI'].values
        X5 = df['temp'].values
        X6 = df['RH'].values
        X7 = df['wind'].values
        X8 = df['rain'].values
        #Create combined X matrix, which will be used to predict Price
        ones = np.ones((268,1))
        X_{all} = np.column_stack((ones, X1, X2, X3, X4, X5, X6, X7, X8))
        lm = LinearRegression()
        lm.fit(X all, Y)
Out[3]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=1, normalize=False)
```

1 Only Ones Column M0 (Null Hypothesis)

```
In [4]: from sklearn.model_selection import KFold # import KFold
    X = ones
    scaler = MinMaxScaler(feature_range=(0,1))
    X = scaler.fit_transform(X)
```

```
kf = KFold(n_splits=5,shuffle=True) # Define the split - into 2 folds
        k_amount = kf.get_n_splits(X) # returns the number of splitting iterations in the cros
In [5]: fig, axes = plt.subplots(3, 2,figsize=(15,15))
        plot = [axes[0,0], axes[0,1], axes[1,0], axes[1,1], axes[2,0], axes[2,1]]
        iter_plot = 0
        total_train_SSE = 0
        total test SSE = 0
        total w = 0
        for train_index, test_index in kf.split(X):
            X_train, X_test = X[train_index], X[test_index]
            Y_train, Y_test = Y[train_index], Y[test_index]
            #Find the weight by using train set
            w_train = np.linalg.lstsq(X_train,Y_train)[0]
            total_w = total_w+w_train
            #Predict the price with train and test set
            pred_train = np.dot(X_train,w_train)
            pred_test = np.dot(X_test,w_train)
            #Calculate SSE for training set and Test set
            train_SSE = sum((pred_train-Y_train)**2)
            test_SSE = sum((pred_test-Y_test)**2)
            total_train_SSE = total_train_SSE + train_SSE
            total_test_SSE = total_test_SSE + test_SSE
            plot[iter_plot].set_title(str(iter_plot+1)+' Fold')
            plot[iter_plot].bar('Train Set',train_SSE, color='k')
            plot[iter_plot].bar('Test Set',test_SSE, color='g')
            plot[iter_plot].set_ylabel('SSE')
            iter_plot = iter_plot+1
        avg_train_SSE = total_train_SSE/k_amount
        avg_test_SSE = total_test_SSE/k_amount
        avg_w = total_w/k_amount
        plot[iter_plot].set_title('AVG of Train SSE vs Test SSE')
        plot[iter_plot].bar('Train Set',avg_train_SSE, color='b')
        plot[iter_plot].bar('Test Set',avg_test_SSE, color='r')
        plot[iter_plot].set_ylabel('SSE')
        print('AVG Train SSE: ',avg_train_SSE,' AVG Test SSE: ',avg_test_SSE)
        print("Model MO: Area = {:.2f}".format(float(avg_w[0])))
AVG Train SSE: 343464.98904 AVG Test SSE: 85866.24726
Model MO: Area = 0.00
```



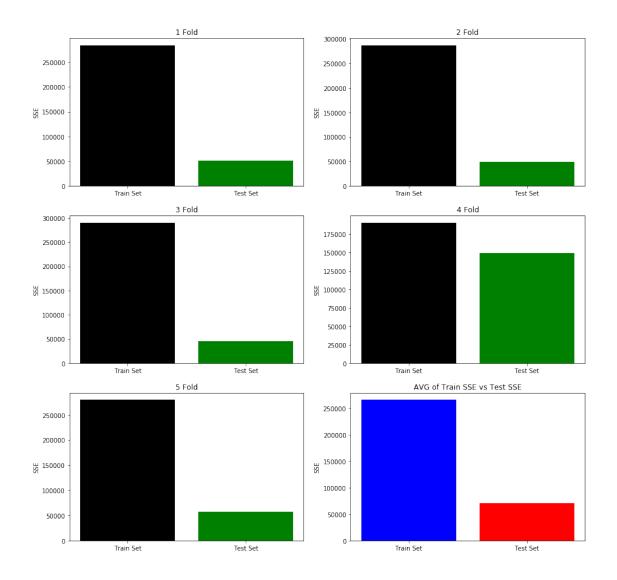
2 With All Variables to Predict Burned Area M1

```
In [6]: from sklearn.model_selection import KFold # import KFold

X = X_all
    scaler = MinMaxScaler(feature_range=(0,1))
X = scaler.fit_transform(X)
    kf = KFold(n_splits=5,shuffle=True) # Define the split - into 2 folds

In [7]: fig, axes = plt.subplots(3, 2,figsize=(15,15))
    plot =[axes[0,0],axes[0,1],axes[1,0],axes[1,1],axes[2,0],axes[2,1]]
    iter_plot = 0
    total_train_SSE = 0
    total_test_SSE = 0
```

```
total_w = 0
                           for train_index, test_index in kf.split(X):
                                         X_train, X_test = X[train_index], X[test_index]
                                        Y_train, Y_test = Y[train_index], Y[test_index]
                                         #Find the weight by using train set
                                         w_train = np.linalg.lstsq(X_train,Y_train)[0]
                                         total_w = total_w+w_train
                                         #Predict the price with train and test set
                                        pred_train = np.dot(X_train,w_train)
                                        pred_test = np.dot(X_test,w_train)
                                         #Calculate SSE for training set and Test set
                                         train_SSE = sum((pred_train-Y_train)**2)
                                         test_SSE = sum((pred_test-Y_test)**2)
                                         total_train_SSE = total_train_SSE + train_SSE
                                         total_test_SSE = total_test_SSE + test_SSE
                                        plot[iter_plot].set_title(str(iter_plot+1)+' Fold')
                                        plot[iter_plot].bar('Train Set',train_SSE, color='k')
                                        plot[iter_plot].bar('Test Set',test_SSE, color='g')
                                        plot[iter_plot].set_ylabel('SSE')
                                         iter_plot = iter_plot+1
                           avg_train_SSE = total_train_SSE/k_amount
                           avg_test_SSE = total_test_SSE/k_amount
                           avg_w = total_w/k_amount
                           plot[iter_plot].set_title('AVG of Train SSE vs Test SSE')
                          plot[iter_plot].bar('Train Set',avg_train_SSE, color='b')
                          plot[iter_plot].bar('Test Set',avg_test_SSE, color='r')
                          plot[iter_plot].set_ylabel('SSE')
                          print('AVG Train SSE: ',avg_train_SSE,' AVG Test SSE: ',avg_test_SSE)
                          print("Model M1: Area = {:.2f} + {:.2f} * FFMC + {:.2f} * DMC + {:.2f} * DC + {:.2f}
                              + \{:.2f\} * temp + \{:.2f\} * RH + \{:.2f\} * wind + \{:.2f\} * rain"
                                                .format(float(avg w[0]), float(avg w[1]), float(avg w[2]), float(avg w[3]), \
                                                                           float(avg_w[4]), float(avg_w[5]),float(avg_w[6]),float(avg_w[7]),float(a
AVG Train SSE: 266232.380103 AVG Test SSE: 70585.8285532
Model M1: Area = 0.00 + 22.11 * FFMC + 17.84 * DMC + -9.19 * DC + -34.00 * ISI + 17.80 * temp = 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.00 + 1.
```



With DMC and DC to predict Burned Area M2 (Highest Positive Correlation)

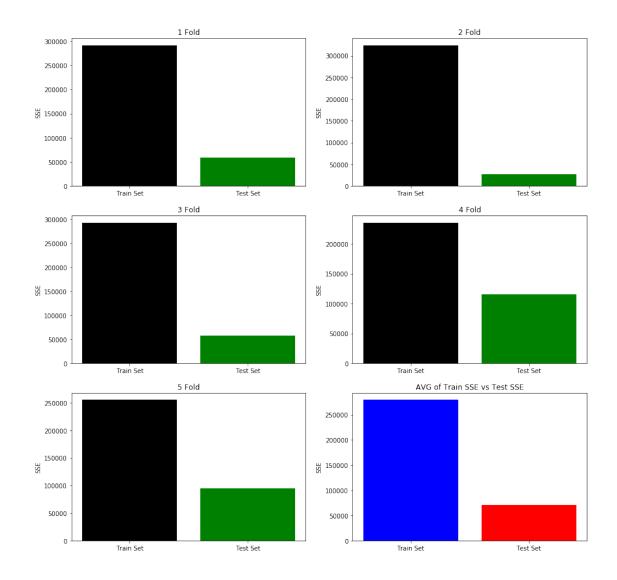
```
In [8]: from sklearn.model_selection import KFold # import KFold

X = np.column_stack((ones, X2, X3)) # create an array
scaler = MinMaxScaler(feature_range=(0,1))

X = scaler.fit_transform(X)
kf = KFold(n_splits=5,shuffle=True) # Define the split - into 2 folds

In [9]: fig, axes = plt.subplots(3, 2,figsize=(15,15))
plot =[axes[0,0],axes[0,1],axes[1,0],axes[1,1],axes[2,0],axes[2,1]]
iter_plot = 0
total_train_SSE = 0
```

```
total_test_SSE = 0
        total_w = 0
        for train_index, test_index in kf.split(X):
            X_train, X_test = X[train_index], X[test_index]
            Y_train, Y_test = Y[train_index], Y[test_index]
            #Find the weight by using train set
            w_train = np.linalg.lstsq(X_train,Y_train)[0]
            total_w = total_w+w_train
            #Predict the price with train and test set
            pred_train = np.dot(X_train,w_train)
            pred_test = np.dot(X_test,w_train)
            #Calculate SSE for training set and Test set
            train_SSE = sum((pred_train-Y_train)**2)
            test_SSE = sum((pred_test-Y_test)**2)
            total_train_SSE = total_train_SSE + train_SSE
            total_test_SSE = total_test_SSE + test_SSE
            plot[iter_plot].set_title(str(iter_plot+1)+' Fold')
            plot[iter_plot].bar('Train Set',train_SSE, color='k')
            plot[iter_plot].bar('Test Set',test_SSE, color='g')
            plot[iter_plot].set_ylabel('SSE')
            iter_plot = iter_plot+1
        avg_train_SSE = total_train_SSE/k_amount
        avg_test_SSE = total_test_SSE/k_amount
        avg_w = total_w/k_amount
        plot[iter_plot].set_title('AVG of Train SSE vs Test SSE')
        plot[iter_plot].bar('Train Set',avg_train_SSE, color='b')
        plot[iter_plot].bar('Test Set',avg_test_SSE, color='r')
        plot[iter_plot].set_ylabel('SSE')
        print('AVG Train SSE: ',avg_train_SSE,' AVG Test SSE: ',avg_test_SSE)
        print()
        print("Model M2: Area = {:.2f} + {:.2f} * DMC + {:.2f} * DC".format(float(avg_w[0]),float(avg_w[0]),float(avg_w[0])
AVG Train SSE: 279631.994948 AVG Test SSE: 71152.7367409
Model M2: Area = 0.00 + 21.66 * DMC + 11.16 * DC
```



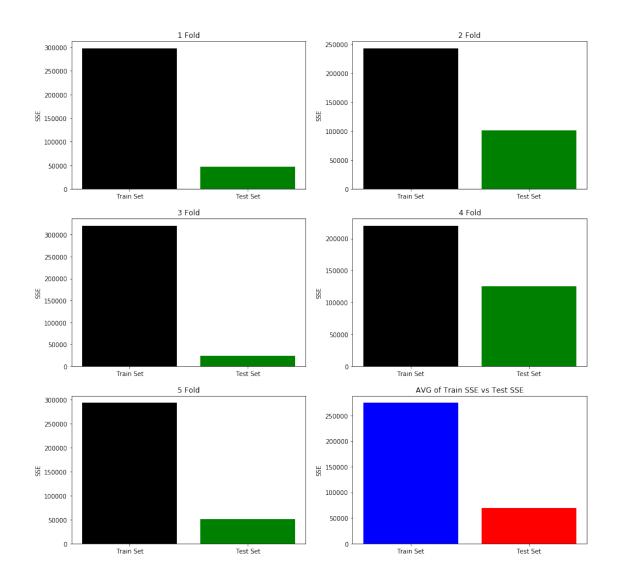
4 With RH and temp to predict Burned Area M3 (Highest Negative Correlation)

```
In [10]: from sklearn.model_selection import KFold # import KFold

X = np.column_stack((ones,X5,X6))# create an array
scaler = MinMaxScaler(feature_range=(0,1))
X = scaler.fit_transform(X)
kf = KFold(n_splits=5,shuffle=True) # Define the split - into 2 folds

In [11]: fig, axes = plt.subplots(3, 2,figsize=(15,15))
plot =[axes[0,0],axes[0,1],axes[1,0],axes[1,1],axes[2,0],axes[2,1]]
iter_plot = 0
total_train_SSE = 0
```

```
total_test_SSE = 0
         total_w = 0
         for train_index, test_index in kf.split(X):
             X_train, X_test = X[train_index], X[test_index]
             Y_train, Y_test = Y[train_index], Y[test_index]
             #Find the weight by using train set
             w_train = np.linalg.lstsq(X_train,Y_train)[0]
             total_w = total_w+w_train
             #Predict the price with train and test set
             pred_train = np.dot(X_train,w_train)
             pred_test = np.dot(X_test,w_train)
             #Calculate SSE for training set and Test set
             train_SSE = sum((pred_train-Y_train)**2)
             test_SSE = sum((pred_test-Y_test)**2)
             total_train_SSE = total_train_SSE + train_SSE
             total_test_SSE = total_test_SSE + test_SSE
             plot[iter_plot].set_title(str(iter_plot+1)+' Fold')
             plot[iter_plot].bar('Train_Set',train_SSE, color='k')
             plot[iter_plot].bar('Test Set',test_SSE, color='g')
             plot[iter_plot].set_ylabel('SSE')
             iter_plot = iter_plot+1
         avg_train_SSE = total_train_SSE/k_amount
         avg_test_SSE = total_test_SSE/k_amount
         avg_w = total_w/k_amount
         plot[iter_plot].set_title('AVG of Train SSE vs Test SSE')
         plot[iter_plot].bar('Train Set',avg_train_SSE, color='b')
         plot[iter_plot].bar('Test Set',avg_test_SSE, color='r')
         plot[iter_plot].set_ylabel('SSE')
         print('AVG Train SSE: ',avg_train_SSE,' AVG Test SSE: ',avg_test_SSE)
         print()
         print("Model M3: Area = {:.2f} + {:.2f} * RH + {:.2f} * temp".format(float(avg_w[0]),
AVG Train SSE: 274840.606274 AVG Test SSE: 69760.2743931
Model M3: Area = 0.00 + 24.77 * RH + 10.48 * temp
```



5 Multipling Correlated Variables to predict Burned Area M4

```
scaler = MinMaxScaler(feature_range=(0,1))
         X = scaler.fit_transform(X)
         y = list(y)
         y = np.array(y)
In [13]: scores = []
         cv = KFold(n_splits=5, shuffle=True)
         k_amount = cv.get_n_splits(X)
         lm = LinearRegression()
         fig, axes = plt.subplots(3, 2,figsize=(15,15))
         plot =[axes[0,0],axes[0,1],axes[1,0],axes[1,1],axes[2,0],axes[2,1]]
         iter_plot = 0
         total_train_SSE = 0
         total_test_SSE = 0
         total_w = 0
         for train_index, test_index in cv.split(X):
             X_train = X[train_index]
             X_test = X[test_index]
             Y_train = y[train_index]
             Y_test = y[test_index]
             lm.fit(X_train, Y_train)
             w=lm.coef
             pred_train = np.dot(X_train,w)
             pred_test = np.dot(X_test,w)
             total_w = total_w+w
             #Calculate SSE for training set and Test set
             train_SSE = sum((pred_train-Y_train)**2)
             test_SSE = sum((pred_test-Y_test)**2)
             total_train_SSE = total_train_SSE + train_SSE
             total_test_SSE = total_test_SSE + test_SSE
             plot[iter_plot].set_title(str(iter_plot+1)+' Fold')
             plot[iter_plot].bar('Train Set',train_SSE, color='k')
             plot[iter_plot].bar('Test Set',test_SSE, color='g')
             plot[iter_plot].set_ylabel('SSE')
             iter_plot = iter_plot+1
             scores.append(lm.score(X_test, Y_test))
         avg_train_SSE = total_train_SSE/k_amount
         avg_test_SSE = total_test_SSE/k_amount
         avg_w = total_w/k_amount
         plot[iter_plot].set_title('AVG of Train SSE vs Test SSE')
         plot[iter_plot].bar('Train Set',avg_train_SSE, color='b')
         plot[iter_plot].bar('Test Set',avg_test_SSE, color='r')
         plot[iter_plot].set_ylabel('SSE')
         print('AVG Train SSE: ',avg_train_SSE,' AVG Test SSE: ',avg_test_SSE)
         print()
         print("Model M4: Area = {:.2f} + {:.2f} * FFMC*ISI + {:.2f} * FFMC*temp + {:.2f} * DM
```

AVG Train SSE: 326904.855862 AVG Test SSE: 85181.9003617

