```
In [1]: import numpy as np
        import pandas as pd
        from sklearn.model selection import train test split
        from sklearn.preprocessing import StandardScaler
        from sklearn.linear model import LinearRegression,LogisticRegression
        from sklearn.neural_network import MLPClassifier,MLPRegressor
        from sklearn.tree import DecisionTreeRegressor as dtr
        from sklearn.ensemble import RandomForestRegressor
        from sklearn.metrics import mean squared error as mse
        from sklearn.metrics import mean_absolute_error as mae
        from sklearn.metrics import r2 score
        import warnings
        warnings.filterwarnings(action='ignore')
In [2]: df=pd.read csv('C:/Users/haris/Downloads/forestfires.csv')
In [3]: df
Out[3]:
```

	X	Y	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area
0	7	5	mar	fri	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.0
1	7	4	oct	tue	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.0
2	7	4	oct	sat	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.0
3	8	6	mar	fri	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.0
4	8	6	mar	sun	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.0
512	4	3	aug	sun	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	6.4
513	2	4	aug	sun	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	54.29
514	7	4	aug	sun	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	11.10
515	1	4	aug	sat	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	0.0
516	6	3	nov	tue	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	0.0

517 rows × 13 columns

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
       RangeIndex: 517 entries, 0 to 516
       Data columns (total 13 columns):
            Column Non-Null Count Dtype
                    _____
        0
           Χ
                    517 non-null
                                    int64
        1
           Υ
                    517 non-null
                                    int64
        2
                    517 non-null
                                   object
           month
        3
                    517 non-null
                                   object
            day
        4
            FFMC
                    517 non-null
                                    float64
        5
            DMC
                    517 non-null
                                    float64
        6
           DC
                    517 non-null
                                   float64
        7
                    517 non-null
                                    float64
            ISI
        8
            temp
                   517 non-null
                                    float64
        9
            RH
                    517 non-null
                                    int64
        10 wind
                    517 non-null
                                   float64
                   517 non-null
        11
           rain
                                   float64
        12 area
                    517 non-null
                                    float64
       dtypes: float64(8), int64(3), object(2)
       memory usage: 52.6+ KB
In [5]: df['month'].unique()
Out[5]: array(['mar', 'oct', 'aug', 'sep', 'apr', 'jun', 'jul', 'feb', 'jan',
               'dec', 'may', 'nov'], dtype=object)
In [6]: df['day'].unique()
Out[6]: array(['fri', 'tue', 'sat', 'sun', 'mon', 'wed', 'thu'], dtype=object)
In [7]: df.head(10)
           X Y month day FFMC
                                             DC
                                                  ISI temp RH wind rain area
Out[7]:
                                     DMC
          7 5
                               86.2
                                            94.3
                                                  5.1
                                                         8.2
                                                                   6.7
                                                                        0.0
                                                                              0.0
        0
                          fri
                                     26.2
                                                              51
                   mar
          7 4
                    oct
                         tue
                               90.6
                                     35.4 669.1
                                                  6.7
                                                        18.0
                                                             33
                                                                   0.9
                                                                        0.0
                                                                              0.0
        1
                                                                        0.0
        2
          7 4
                    oct
                         sat
                               90.6
                                     43.7 686.9
                                                  6.7
                                                        14.6
                                                             33
                                                                   1.3
                                                                              0.0
        3 8 6
                          fri
                               91.7
                                     33.3
                                            77.5
                                                  9.0
                                                         8.3
                                                             97
                                                                   4.0
                                                                        0.2
                                                                              0.0
                   mar
                               89.3
                                     51.3 102.2
                                                  9.6
                                                        11.4
                                                              99
                                                                   1.8
                                                                        0.0
                                                                              0.0
        4
           8 6
                   mar
                         sun
                                     85.3 488.0 14.7
                                                                        0.0
                                                                              0.0
        5 8 6
                               92.3
                                                        22.2
                                                             29
                                                                   5.4
                   aug
                         sun
        6 8 6
                   aug
                        mon
                               92.3
                                     88.9 495.6
                                                  8.5
                                                        24.1
                                                             27
                                                                   3.1
                                                                        0.0
                                                                              0.0
                               91.5 145.4 608.2 10.7
                                                         8.0 86
                                                                   2.2
                                                                        0.0
                                                                              0.0
        7 8 6
                   aug
                        mon
        8 8 6
                    sep
                         tue
                               91.0 129.5 692.6
                                                  7.0
                                                        13.1
                                                             63
                                                                   5.4
                                                                        0.0
                                                                              0.0
        9 7 5
                    sep
                         sat
                               92.5
                                     88.0 698.6
                                                  7.1
                                                        22.8
                                                             40
                                                                   4.0
                                                                        0.0
                                                                              0.0
```

Preprocessing

Text data included with the numeric data(Month & Days). So we need to encode that in some numeric form before splitting the train test data.

```
In [8]: def ordinal encoding(df,column,ordering):
          df=df.copy()
          df[column]=df[column].apply(lambda x: ordering.index(x))
          return df
In [9]: def preprocessing(df,task):
          df=df.copy()
          df=ordinal encoding(
              df,
              column='month',
              ordering=[
                          'jan', 'feb', 'mar', 'apr', 'may', 'jun', 'jul', 'aug', 'se
               'oct', 'nov', 'dec'
          )
          df=ordinal encoding(
              df,
              column='day',
              ordering=['mon', 'tue', 'wed', 'thu', 'fri', 'sat', 'sun']
          if task=='Regression':
            Y=df['area']
          elif task=='Classification':
            Y=df['area'].apply(lambda x: 1 if x>0 else 0)
          X=df.drop('area',axis=1)
          X_train,X_test,Y_train,Y_test=train_test_split(X,Y,train_size=0.65,shuffle
          scaler=StandardScaler()
          scaler.fit(X train)
          X train=pd.DataFrame(scaler.transform(X train),columns=X.columns)
          X test=pd.DataFrame(scaler.transform(X test),columns=X.columns)
          return X train,X test,Y train,Y test
```

Splitting & Testing Models:

```
In [10]: X train,X test,Y train,Y test=preprocessing(df,task='Regression')
In [11]: X train.head()
                            Υ
                                                      FFMC
                                                                 DMC
                                                                            DC
Out[11]:
                  X
                                 month
                                              day
            1.460144
                      1.330887 0.226770 -0.599195 -0.750425 -0.262938
                                                                      0.268405 -0
            0.603934 -0.246461 0.226770 -0.109342
                                                   0.748288 0.383050
                                                                      0.145783
                                                                                0
         2 -0.252276  0.542213 -1.913533  0.380511
                                                   0.201882 -1.156059 -1.850008
                                                                                0
         3 -1.108486 0.542213 0.226770 -1.089047
                                                   0.264329 0.709954
                                                                      0.461893
                                                                               1
         4 1.460144 1.330887 -1.913533 0.380511
                                                   0.201882 -1.116955 -1.836870 -0
```

```
In [12]: X test.head()
Out[12]:
                   Χ
                             Υ
                                  month
                                               day
                                                       FFMC
                                                                  DMC
                                                                              DC
         0 -1.108486 -1.823808 0.226770 -1.089047
                                                     0.264329
                                                               0.709954 0.461893
                                                                                  1.1
            0.603934 0.542213 0.226770
                                          0.870363 -0.032292 -0.161269 0.326531 -0.0
         2 -0.252276 1.330887 0.654831 -0.109342 0.514114 -0.411530 0.569386
                                                                                 1.8
         3 -0.680381 -0.246461 0.654831 0.380511
                                                    0.451668
                                                              0.531643 0.683647 1.0
           1.032039 -0.246461 0.654831 -0.599195 -0.047903 -0.380248 0.770438 -0.5
In [13]: Y train.head()
Out[13]: 171
                2.69
         161
                1.90
         69
                0.00
         272
                3.09
         91
                0.00
         Name: area, dtype: float64
In [14]: Y test.head()
Out[14]: 270
                 0.52
         90
                 0.00
         133
                 0.00
         221
                35.88
         224
                37.71
         Name: area, dtype: float64
         Linear Regression
In [15]: \#The maximum val. of R^2 can be 1.0 that signifies that Linear Regression is
         #Here R^2 score is 0.02051 which is really low signifying that LinearRegress
         linear_reg_model=LinearRegression()
         linear reg model.fit(X train,Y train)
         print("Performance of Linear Regression R^2 metric {:.5f}".format(linear reg
        Performance of Linear Regression R^2 metric 0.02051
         MLP Regressor Model
In [16]: # We can see the MLP Regressor Model Works better than Linear Regression, how
         mlp reg model=MLPRegressor().fit(X train,Y train)
         print("Performance of MLP Regressor Model R^2 metric {:.5f}".format(mlp reg
        Performance of MLP Regressor Model R^2 metric 0.06170
         Decision Tree Regressor
In [17]: #R2 score negative not a good fit!
```

reg = dtr(random_state = 42)
reg.fit(X train, Y train)

```
Y_pred = reg.predict(X_test)
print("MSE =", mse(Y_pred, Y_test))
print("MAE =", mae(Y_pred, Y_test))
print("R2 Score =", r2_score(Y_pred, Y_test))

MSE = 11504.043293370165
MAE = 27.108121546961325
R2 Score = -7.943021164352407
```

Random Forest Regressor

```
In [18]: #This works even worse than Decision Tree Regressor
    regr = RandomForestRegressor(max_depth=2, random_state=0, n_estimators=100)
    regr.fit(X_train, Y_train)
    Y_pred = regr.predict(X_test)
    print("MSE =", mse(Y_pred, Y_test))
    print("MAE =", mae(Y_pred, Y_test))
    print("R2 Score =", r2_score(Y_pred, Y_test))

MSE = 9900.080101011921
    MAE = 22.488793674759876
    R2 Score = -588.768145917561
```

Reaction after seeing the R^2 Score everytime getting less and less!!!!!!

Logistic Regression

```
In [19]: X train,X test,Y train,Y test=preprocessing(df,task='Classification')
In [20]: Y train
Out[20]: 171
                1
         161
                1
         69
                0
         272
                1
         91
                0
         129
              0
         144
                1
         72
                0
         235
                1
         37
         Name: area, Length: 336, dtype: int64
In [21]: log reg model=LogisticRegression()
         log reg model.fit(X train,Y train)
         print('Logistic Regression Accuracy, {:.5f}%'.format(log_reg_model.score(X_t
        Logistic Regression Accuracy, 53.59116%
```

Neural Network Classifier

```
In [22]: nn_classifier_model=MLPClassifier(activation='relu', hidden_layer_sizes=(16,1 nn_classifier_model.fit(X_train,Y_train)
```

```
Out[22]: MLPClassifier

MLPClassifier(hidden_layer_sizes=(16, 16), n_iter_no_change=100)
```

In [23]: print('MLP Classifier Accuracy, {:.5f}%'.format(nn_classifier_model.score(X_
MLP Classifier Accuracy, 55.80110%

This notebook was converted with convert.ploomber.io