Model Building

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Project Name	DEMANDEST – AI POWERED FOOD
	DEMAND FORECASTER

Training and testing algorithmic models:

Data can be processed using a variety of Machine Learning models, including images, sound, text and numerical values. Algorithms can be chosen according to the objective. They can be either Classification models or Regression models.

- Linear Regression Model
- Lasso Regression model
- Elastic Net Regression model
- XGB Regressor model
- Gradient Boosting Model
- Decision Tree
- K-Nearest Neighbour

```
In [85]: # Split the dataset into train and test data

X = final_train3.values
Y = final_train['num_orders'].values
X_train,X_val,Y_train,Y_val = train_test_split(X,Y,test_size=0.25)
```

Model evaluation:

In Machine Learning, mathematical models are constructed to make sense of the data at hand. By fitting these models to previously observed data, new data can be predicted.

```
In [86]: # Linear Regression Model
         LR_mod = LinearRegression()
         LR_mod.fit(X_train,Y_train)
         Y_pred = LR_mod.predict(X_val)
         Y_pred[Y_pred<0] = 0
         # Root Mean Square Error
         RMSE = 100*np.sqrt(metrics.mean_squared_log_error(Y_val,Y_pred))
         RMSE
Out[86]: 129.70651046086593
In [87]: # Lasso Regression model
         L_{mod} = Lasso()
         L_mod.fit(X_train,Y_train)
         Y_pred = L_mod.predict(X_val)
         Y_pred[Y_pred<0] = 0
         # Root Mean Square Error
         RMSE = 100*np.sqrt(metrics.mean_squared_log_error(Y_val,Y_pred))
         RMSE
Out[87]: 129.20484156819074
 In [90]: # Decision Tree
           Dec_tree = DecisionTreeRegressor()
           Dec_tree.fit(X_train,Y_train)
           Y_pred = Dec_tree.predict(X_val)
           Y_pred[Y_pred<0] = 0
           # Root Mean Square Error
           RMSE = 100*np.sqrt(metrics.mean_squared_log_error(Y_val,Y_pred))
           RMSE
 Out[90]: 62.89494154472869
 In [96]: # K-Nearest Neighbor
           KNN = KNeighborsRegressor()
           KNN.fit(X_train,Y_train)
           Y_pred = KNN.predict(X_val)
           Y_pred[Y_pred<0] = 0
           # Root Mean Square Error
           RMSE = 100*np.sqrt(metrics.mean_squared_log_error(Y_val,Y_pred))
           RMSE
 Out[96]: 67.24720843560212
```

Save the model:

To test the machine learning model on new data, save the trained model to a file and restore them.

```
In [109]: # To store the Decision Tree model
pickle.dump(Dec_tree,open('demand_forecast.pkl','wb'))
```

Using the model to predict the output:

```
In [114]: # Predicting the number of orders using Decision Tree Model
          final test = pd.merge(test data,food data,on="meal id",how="outer")
          final test = pd.merge(final test,centers data,on="center id",how="outer")
          final test = final test.drop(['center id','meal id'],axis=1)
          fcolumns = final test.columns.tolist()
          fcolumns = fcolumns[:2]+fcolumns[8:]+fcolumns[6:8]+fcolumns[2:6]
          final_test = final_test[fcolumns]
          le = LabelEncoder()
          final test['center type'] = le.fit transform(final test['center type'])
          final_test['category'] = le.fit_transform(final_test['category'])
          final_test['cuisine'] = le.fit_transform(final_test['cuisine'])
          X test = final test[features].values
          obser = Dec tree.predict(X test)
          obser[obser<0] = 0
          result = pd.DataFrame({'id':final test['id'], 'num orders':obser})
          result.to csv("C:/Users/91948/Downloads/IBM PROJECT/Datasets/submission.csv",index=False)
          result.describe()
```

```
Out[114]: id num_orders

count 3.257300e+04 32573.000000

mean 1.248476e+06 262.516689
```