

TEAM ID: PNT2022TMID52731

PROJECT NAME: DemandEst - AI powered Food Demand Forecaster

Team Leader

The screenshot displays a Jupyter Notebook interface with the following content:

Model Evaluation

We're going to use `x_train` and `y_train` obtained above in `train_test_split` section to train our regression model. We're using the `fit` method and passing the parameters as shown below. Finally, we need to check to see how well our model is performing on the test data.

Regression Evaluation Metrics: RMSE: Root Mean Square Error RMSE is the square root of the averaged squared difference between the target value and the value predicted by the model. It is preferred more in some cases because the errors are first squared before averaging which poses a high penalty on large errors. This implies that RMSE is useful when large errors are undesired.

For testing the model we use the below method,

```
In [126]: XG = XGBRegressor()
XG.fit(X_train, y_train)
y_pred = XG.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 70.06429878638917
```

```
In [127]: L = Lasso()
L.fit(X_train, y_train)
y_pred = L.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 128.9558620089095
```

```
In [128]: EN = ElasticNet()
EN.fit(X_train, y_train)
y_pred = EN.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 130.93230794494932
```

```
In [129]: DT = DecisionTreeRegressor()
DT.fit(X_train, y_train)
y_pred = DT.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 62.750116693228705
```

```
In [130]: KNN = KNeighborsRegressor()
KNN.fit(X_train, y_train)
y_pred = KNN.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152
```

```
In [131]: GB = GradientBoostingRegressor()
GB.fit(X_train, y_train)
```

```
RMSLE: 130.93230794494932

In [129]: DT = DecisionTreeRegressor()
DT.fit(X_train, y_train)
y_pred = DT.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 62.750116693228705

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y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152

In [131]: GB = GradientBoostingRegressor()
GB.fit(X_train, y_train)
y_pred = GB.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 99.04866931366767
```

Team Member 1

```
Model Evaluation

We're going to use x_train and y_train obtained above in train_test_split section to train our regression model. We're using the fit method and passing the parameters as shown below. Finally, we need to check to see how well our model is performing on the test data.

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y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 70.06429878638917

In [127]: L = Lasso()
L.fit(X_train, y_train)
y_pred = L.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 128.9558620089095
```

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File Edit View Insert Cell Kernel Widgets Help Not Connected Not Trusted Python 3 (ipykernel)

Run

```
In [128]: EN = ElasticNet()
          EN.fit(X_train, y_train)
          y_pred = EN.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 130.93230794494932
```

```
In [129]: DT = DecisionTreeRegressor()
          DT.fit(X_train, y_train)
          y_pred = DT.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 62.750116693228705
```

```
In [130]: KNN = KNeighborsRegressor()
          KNN.fit(X_train, y_train)
          y_pred = KNN.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152
```

```
In [131]: GB = GradientBoostingRegressor()
          GB.fit(X_train, y_train)
```

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Gmail YouTube Maps

jupyter Code (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Not Connected Not Trusted Python 3 (ipykernel)

Run

```
RMSLE: 130.93230794494932
```

```
In [129]: DT = DecisionTreeRegressor()
          DT.fit(X_train, y_train)
          y_pred = DT.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 62.750116693228705
```

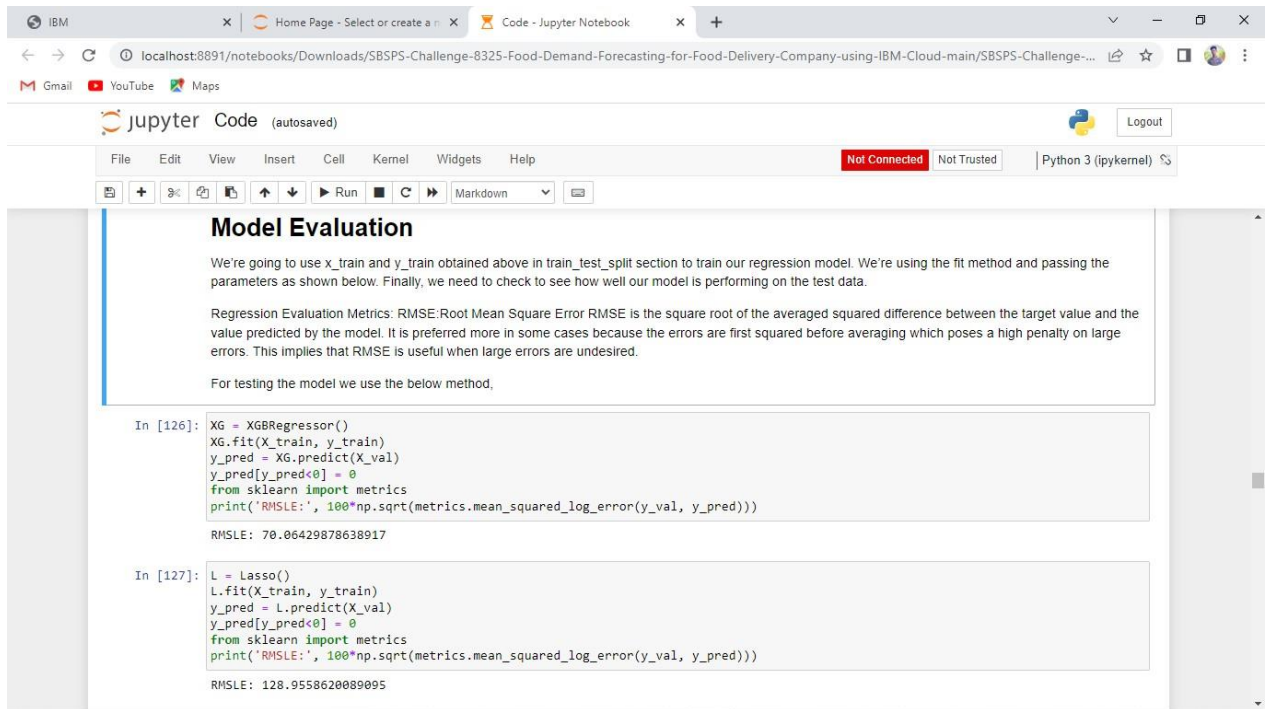
```
In [130]: KNN = KNeighborsRegressor()
          KNN.fit(X_train, y_train)
          y_pred = KNN.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152
```

```
In [131]: GB = GradientBoostingRegressor()
          GB.fit(X_train, y_train)
          y_pred = GB.predict(X_val)
          y_pred[y_pred<0] = 0
          from sklearn import metrics
          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 99.04866931366767
```

Team Member 2



The screenshot shows a Jupyter Notebook interface with a browser window at the top. The notebook has a title bar with 'jupyter Code (autosaved)' and a 'Logout' button. The interface includes a menu bar (File, Edit, View, Insert, Cell, Kernel, Widgets, Help) and a toolbar with icons for file operations, running, and markdown. The main content area is titled 'Model Evaluation' and contains the following text:

We're going to use `x_train` and `y_train` obtained above in `train_test_split` section to train our regression model. We're using the `fit` method and passing the parameters as shown below. Finally, we need to check to see how well our model is performing on the test data.

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For testing the model we use the below method,

In [126]:

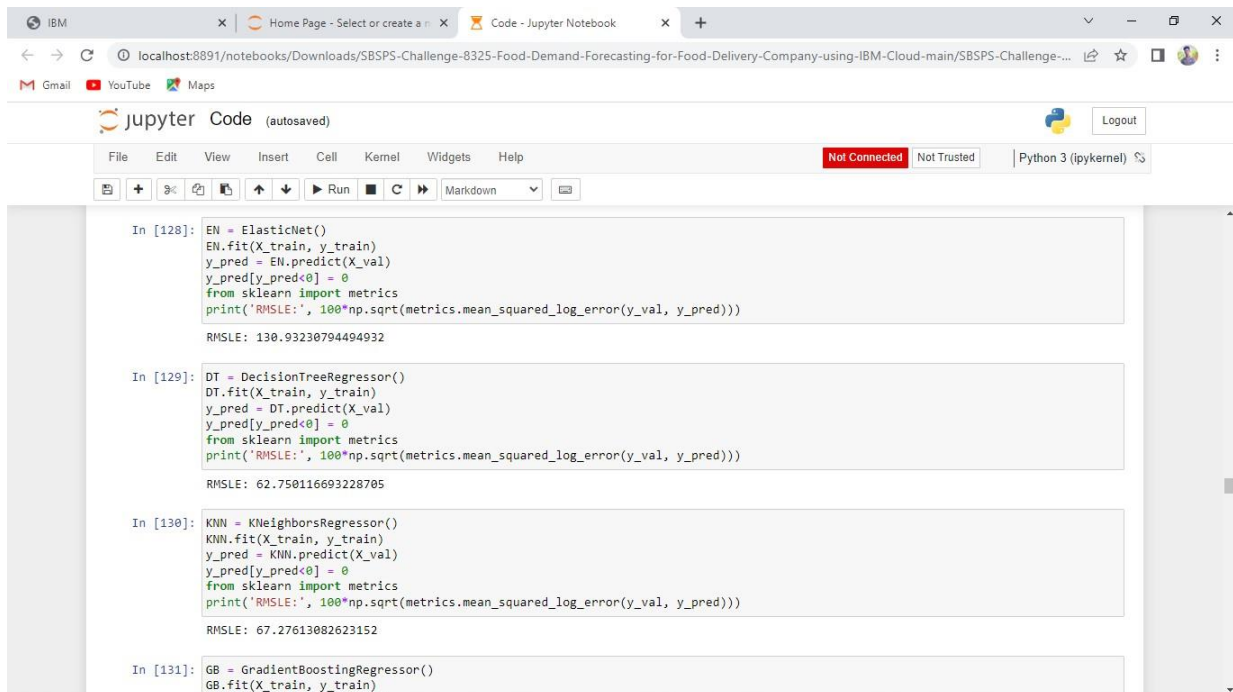
```
XG = XGBRegressor()
XG.fit(X_train, y_train)
y_pred = XG.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
```

RMSLE: 70.06429878638917

In [127]:

```
L = Lasso()
L.fit(X_train, y_train)
y_pred = L.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
```

RMSLE: 128.9558620809095



The screenshot shows a Jupyter Notebook interface with a browser window at the top. The notebook has a title bar with 'jupyter Code (autosaved)' and a 'Logout' button. The interface includes a menu bar (File, Edit, View, Insert, Cell, Kernel, Widgets, Help) and a toolbar with icons for file operations, running, and markdown. The main content area contains four code cells, each testing a different regression model:

In [128]:

```
EN = ElasticNet()
EN.fit(X_train, y_train)
y_pred = EN.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
```

RMSLE: 130.93230794494932

In [129]:

```
DT = DecisionTreeRegressor()
DT.fit(X_train, y_train)
y_pred = DT.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
```

RMSLE: 62.750116693228705

In [130]:

```
KNN = KNeighborsRegressor()
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y_pred = KNN.predict(X_val)
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from sklearn import metrics
print('RMSE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
```

RMSLE: 67.27613082623152

In [131]:

```
GB = GradientBoostingRegressor()
GB.fit(X_train, y_train)
```

```
RMSLE: 130.93230794494932

In [129]: DT = DecisionTreeRegressor()
DT.fit(X_train, y_train)
y_pred = DT.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 62.750116693228705

In [130]: KNN = KNeighborsRegressor()
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y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152

In [131]: GB = GradientBoostingRegressor()
GB.fit(X_train, y_train)
y_pred = GB.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 99.04866931366767
```

Team Member 3

```
Model Evaluation

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print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 70.06429878638917

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print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 128.9558620089095
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jupyter Code (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Not Connected Not Trusted Python 3 (ipykernel)

Run

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          from sklearn import metrics
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          print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))

RMSLE: 67.27613082623152
```

```
In [131]: GB = GradientBoostingRegressor()
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```

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jupyter Code (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Not Connected Not Trusted Python 3 (ipykernel)

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RMSLE: 130.93230794494932
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RMSLE: 99.04866931366767
```