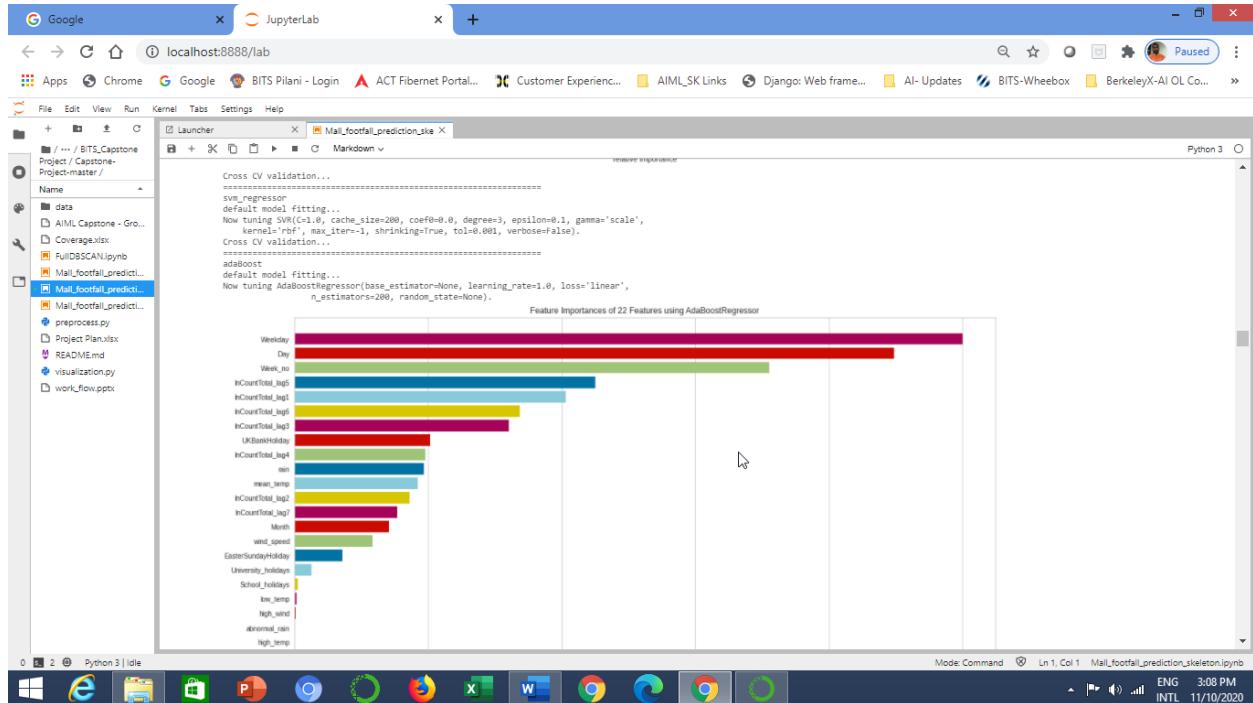
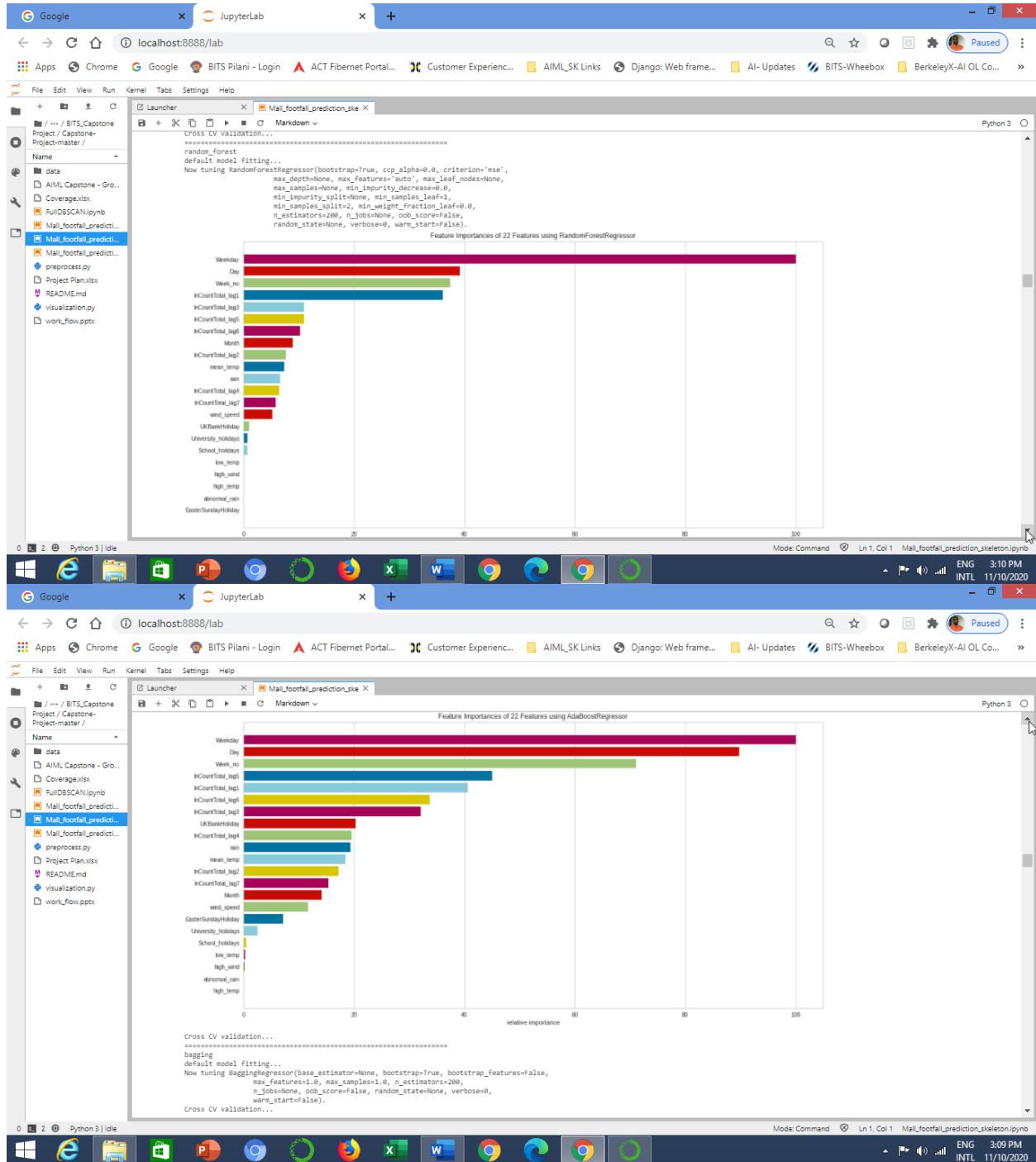


## Code SNIPPETS PART-2





```

Google JupyterLab
localhost:8888/lab
File Edit View Run Kernel Tabs Settings Help
Launcher Mail_footfall_prediction_ske Python 3
random_forest
default model fitting...
Now tuning RandomForestRegressor('bootstrap=True, ccp_alpha=0.0, criterion='mse',
max_depth=None, max_features='auto', max_leaf_nodes=None,
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_leaf=0.0,
n_estimators=200, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False).

Feature Importances of 22 Features using RandomForestRegressor
Weekday Day Week_2o Week_no
ICountTotal_lag1 ICountTotal_lag2 ICountTotal_lag3
ICountTotal_lag4 ICountTotal_lag5
ICountTotal_lag6 Month
ICountTotal_lag7 mean_temp
ICountTotal_lag8 rain
ICountTotal_lag9 wind_speed
UKBankHoliday University_holidays
School_holidays
Extemp Nhigh_wend Nhigh_temp
abnormal_rain EasterSundayHoliday
relative importance
0 20 40 60 80 100
Mode: Command Ln 1, Col 1 Mail_footfall_prediction_skeleton.ipynb
0 2 3:12 PM INTL 11/10/2020

```

```

Google JupyterLab
localhost:8888/lab
File Edit View Run Kernel Tabs Settings Help
Launcher Mail_footfall_prediction_ske Python 3
Cross CV Validation...
extra_trees_regressor
default model fitting...
Now tuning ExtraTreesRegressor('bootstrap=False, ccp_alpha=0.0, criterion='mse',
max_depth=None, max_features='auto', max_leaf_nodes=None,
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_leaf=0.0,
n_estimators=200, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False).

Feature Importances of 22 Features using ExtraTreesRegressor
Weekday Day Week_2o Week_no
ICountTotal_lag1 ICountTotal_lag2 ICountTotal_lag3
ICountTotal_lag4 ICountTotal_lag5
ICountTotal_lag6 Month
ICountTotal_lag7 wind_speed
ICountTotal_lag8 rain
ICountTotal_lag9 mean_temp
UKBankHoliday University_holidays
School_holidays
Extemp Nhigh_wend Nhigh_temp
abnormal_rain EasterSundayHoliday
relative importance
0 20 40 60 80 100
Mode: Command Ln 1, Col 1 Mail_footfall_prediction_skeleton.ipynb
0 2 3:13 PM INTL 11/10/2020

```

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```
Cross CV validation...
=====
gradient_boosting_regressor
default model fitting...
Now tuning GradientBoostingRegressor(base_score=0.9, ccp_alpha=0.0, criterion='friedman_mse',
           init=None, learning_rate=1, loss='ls', max_depth=3,
           max_features=None, max_leaf_nodes=None,
           min_impurity_decrease=0.0, min_impurity_split=None,
           min_samples_leaf=1, min_samples_split=2,
           min_weight_fraction_leaf=0.0, n_estimators=100,
           n_iter_no_change=10, presort='auto',
           random_state=None, subsample=1.0, tol=0.001,
           validation_fraction=0.1, verbose=0, warm_start=False).
```

Feature Importances of 22 Features using GradientBoostingRegressor

Feature	Relative Importance
Weekday	~95
Day	~45
Week_no	~35
InCountTotal_1stP	~30
InCountTotal_2ndP	~28
InCountTotal_3rdP	~25
InCountTotal_4thP	~22
InCountTotal_5thP	~20
Month	~18
Sun	~15
InCountTotal_6thP	~12
mean_temp	~10
wind_speed	~8
UKBankHoliday	~5
University_Holidays	~3
School_Holidays	~2
hgt_temp	~1
low_temp	~1
high_wind	~1
abnormal_rain	~1
EasterSundayHoliday	~1

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```
XGBBoost
default model fitting...
Now tuning XGBRegressor(base_score=0.5, booster='gbtree', colsample_bytree=1,
           colsample_node=1, colsample_bylevel=1, gamma=0,
           importance_type='gain', learning_rate=0.1, max_delta_step=0,
           max_depth=3, min_child_weight=1, missing=None, n_estimators=100,
           n_jobs=1, nthread=None, objective='reg:squarederror',
           random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
           seed=None, silent=None, subsample=1, verbosity=1).
```

Feature Importances of 22 Features using XGBRegressor

Feature	Relative Importance
hgt_wind	~95
abnormal_rain	~45
EasterSundayHoliday	~35

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3

The figure shows a Jupyter Notebook interface with a Python 3 kernel. The main content area displays a horizontal bar chart titled 'relative importance' for various features in a model named 'Mail\_footfall\_prediction\_ske'. The x-axis ranges from 0 to 100. The most important feature is 'mean\_temp', followed by 'wind\_speed', 'School\_holidays', 'low\_temp', 'high\_temp', 'high\_wind', 'abnormal\_rain', and 'EasterSundayHoliday'. Below the chart, the notebook displays several lines of code related to cross-validation and different regression models (LightGBM, CatBoost, and NeuralNet\_MLP) with their respective parameters.

```
mean_temp  
wind_speed  
School_holidays  
low_temp  
high_temp  
high_wind  
abnormal_rain  
EasterSundayHoliday
```

Cross CV validation...  
\*\*\*\*\*  
LightGradientBoosting  
default model fitting...  
Now tuning LightGradientBooster(boosting\_type='gbdt', class\_weight=None, colsample\_bytree=1.0,  
 importance\_type='split', learning\_rate=0.1, max\_depth=3,  
 min\_child\_samples=20, min\_child\_weight=0.01, min\_split\_gain=0.0,  
 n\_estimators=100, n\_jobs=-1, num\_leaves=31, objective=None,  
 random\_state=None, reg\_alpha=0.0, reg\_lambda=0.0, silent=True,  
 subsample=1.0, subsample\_for\_bin=200000, subsample\_freq=0).  
Cross CV validation...  
\*\*\*\*\*  
CategoricalGradientBoosting  
default model fitting...  
Now tuning CategoricalGradientBooster object at 0x7f927875c900.  
Cross CV validation...  
\*\*\*\*\*  
NeuralNet\_MLP  
default model fitting...  
Now tuning MLPRegressor(activation='relu', alpha=0.0001, batch\_size='auto', beta\_1=0.9,  
 beta\_2=0.999, early\_stopping=False, epsilon=1e-09,  
 hidden\_layer\_sizes=(100,), learning\_rate='constant',  
 learning\_rate\_init=0.001, max\_fun=10000, max\_iter=200,  
 momentum=0.9, n\_iter\_no\_change=10, nesterovs\_momentum=True,  
 power\_t=0.5, random\_state=None, shuffle=True, solver='adam',  
 tol=0.0001, validation\_fraction=0.1, verbose=False,  
 warm\_start=False).  
Cross CV validation...  
\*\*\*\*\*

	Model_type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
27	extra_tree_regressor_normal_default_model	17.897495	0.009280	-0.019244	0.068120	-0.025908	0.096229	-0.008740	-0.078412	-1625.491227	-1536.393304
45	gradient_boosting_regressor_normal_default_model	18.208099	0.005594	-0.019348	0.052209	0.026533	0.077495	0.390568	0.348475	-1805.398885	-1716.205962
54	CategoricalGradientBoosting_regressor_normal_default_model	19.219415	0.005014	-0.023464	0.048502	0.057750	0.070226	0.453540	0.415797	-1844.330758	-1755.142033
39	random_forest_normal_default_model	19.436740	0.004790	-0.011844	0.047131	0.066519	0.092113	0.478152	0.442109	-1880.783347	-1771.595425
36	bagging_regressor_normal_default_model	19.531226	0.004843	-0.012142	0.047241	0.066851	0.095954	0.472391	0.435949	-1858.653338	-1767.675415
48	XGBoost_normal_default_model	19.618281	0.005520	-0.017470	0.052320	0.044728	0.074395	0.398708	0.357175	-1810.193338	-1721.004398
51	LightGradientBoosting_normal_default_model	20.169017	0.005385	-0.015131	0.050980	0.050309	0.073248	0.415568	0.375202	-1620.347930	-1731.160007
52	LightGradientBoosting_normal_rcv1_model	20.187940	0.005135	-0.015762	0.049409	0.059171	0.071660	0.440596	0.401959	-1835.973252	-1745.853520
43	extra_tree_regressor_normal_rcv1_model	20.410401	0.005220	-0.018227	0.049977	0.059561	0.072247	0.431407	0.392135	-1830.156794	-1740.988687
55	CategoricalGradientBoosting_normal_rcv1_model	20.421376	0.005221	-0.016520	0.048908	0.060150	0.072355	0.431278	0.391997	-1830.075598	-1740.887073
46	gradient_boosting_regressor_normal_rcv1_model	20.449415	0.005591	-0.016655	0.052405	0.055558	0.074778	0.390899	0.348029	-1805.582248	-1716.400325
42	extra_tree_regressor_normal_default_model	20.531123	0.005240	-0.014965	0.050912	0.052613	0.072445	0.428278	0.388790	-1828.197433	-1730.095910
34	adaBoost_normal_rcv1_model	20.591988	0.005287	-0.009772	0.051323	0.084494	0.072712	0.424054	0.384274	-1825.569446	-1734.381545
58	NeuralNet_MLP_normal_rcv1_model	21.262401	0.004108	-0.017769	0.056421	0.042098	0.078140	0.334887	0.288927	-1774.171149	-1684.993226
40	random_forest_regressor_normal_rcv1_model	21.409627	0.005175	-0.014287	0.048761	0.076737	0.071938	0.436251	0.397313	-1833.210883	-1740.229207
7	ridgeRegressor_normal_rcv1_model	22.012285	0.005185	-0.019548	0.055874	0.055486	0.078653	0.328091	0.279545	-1769.491341	-1680.303418
1	linear_regression_normal_rcv1_model	22.037022	0.005208	-0.019980	0.055071	0.054719	0.070792	0.323709	0.276998	-1768.231746	-1679.043023
0	linear_regression_normal_default_model	22.037022	0.005208	-0.019980	0.055071	0.054719	0.070792	0.323709	0.276998	-1768.231745	-1679.043023
33	adaBoost_normal_default_model	22.278525	0.005211	0.004033	0.059023	0.111025	0.078811	0.323387	0.276054	-1768.061682	-1678.737375
37	bagging_normal_rcv1_model	22.346926	0.005629	-0.049740	0.051728	0.083197	0.057630	0.386757	0.344401	-1803.168424	-1713.805051
49	XGBoost_normal_rcv1_model	22.492584	0.005520	-0.009995	0.051373	0.057399	0.074299	0.396642	0.357107	-1810.155363	-1720.967440
13	huberRegressor_normal_rcv1_model	22.570989	0.005935	-0.020316	0.054370	0.060713	0.079784	0.306584	0.238690	-1759.304395	-1670.116473
10	elastic_net_normal_rcv1_model	22.617232	0.005085	-0.017668	0.054756	0.048831	0.078005	0.337158	0.291376	-1775.402359	-1688.214666

Google JupyterLab

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File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

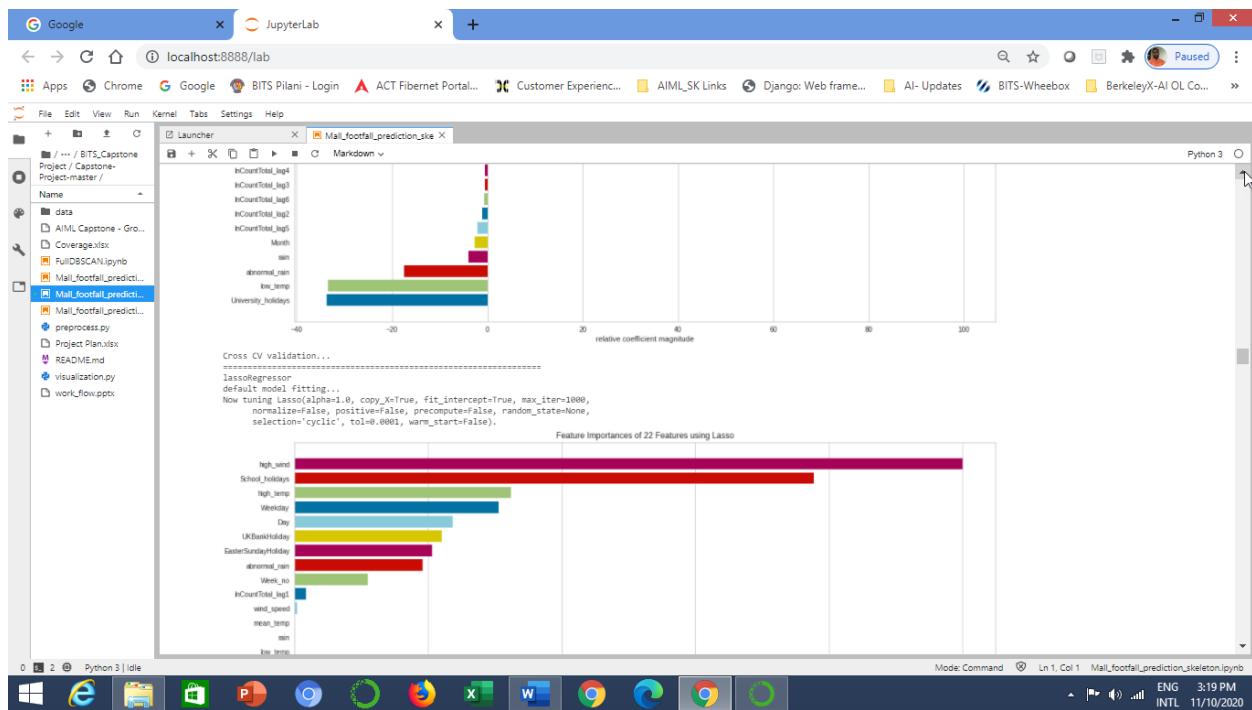
```
[49]: XGBBoost_normal_rscv_model 22.492384 -0.005320 -0.009695 -0.051373 0.057539 0.074299 -0.038842 -0.037107 -1810.155363 -1720.987440
[13]: huberRegressor_normal_rscv_model 22.570989 0.006365 -0.020316 0.094370 0.060713 0.079784 0.306584 0.258690 -1759.304396 -1670.116473
[10]: elasticNet_normal_rscv_model 22.612732 0.006085 -0.017666 0.054756 0.068531 0.078005 0.337158 0.291376 -1775.402589 -1686.214666
[4]: lassoRegressor_normal_rscv_model 22.625361 0.006093 -0.017766 0.054764 0.068460 0.078056 0.334287 0.290445 -1774.934249 -1685.745326
[56]: CategoricalGradientBoosting_normal_rscv_model 22.929772 0.005290 0.001239 0.049439 13.012325 0.072474 0.405185 NaN NaN NaN
```

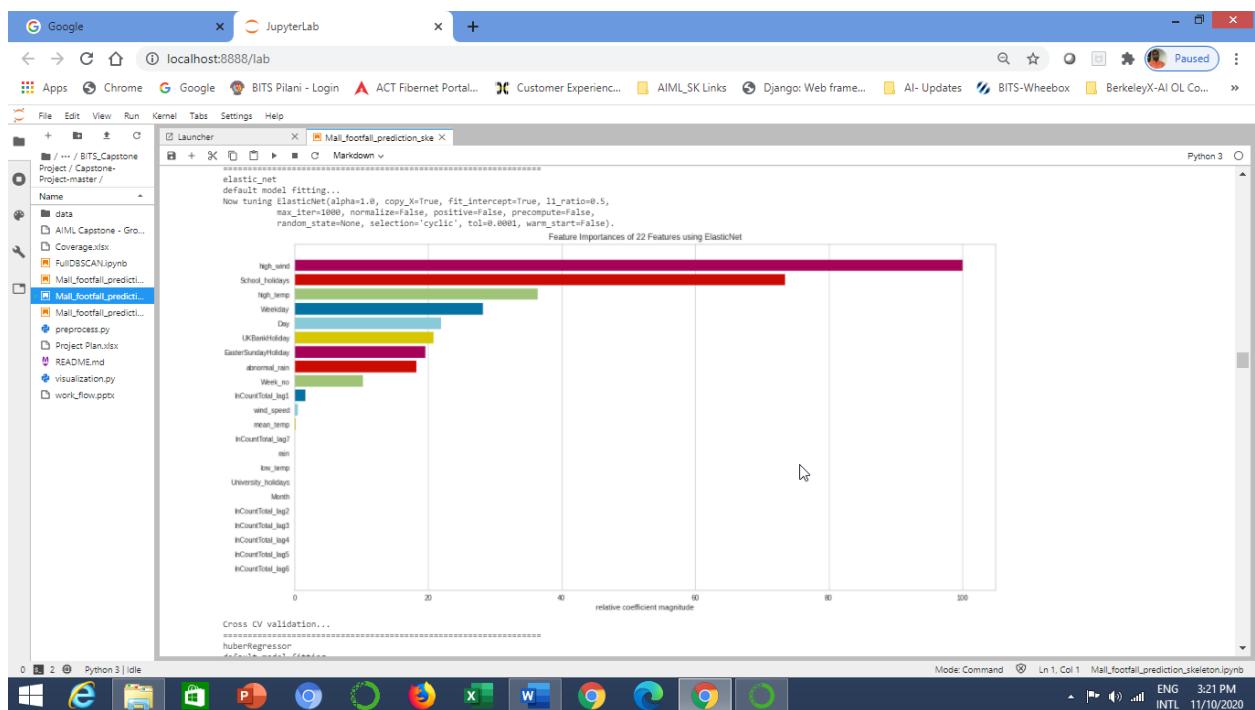
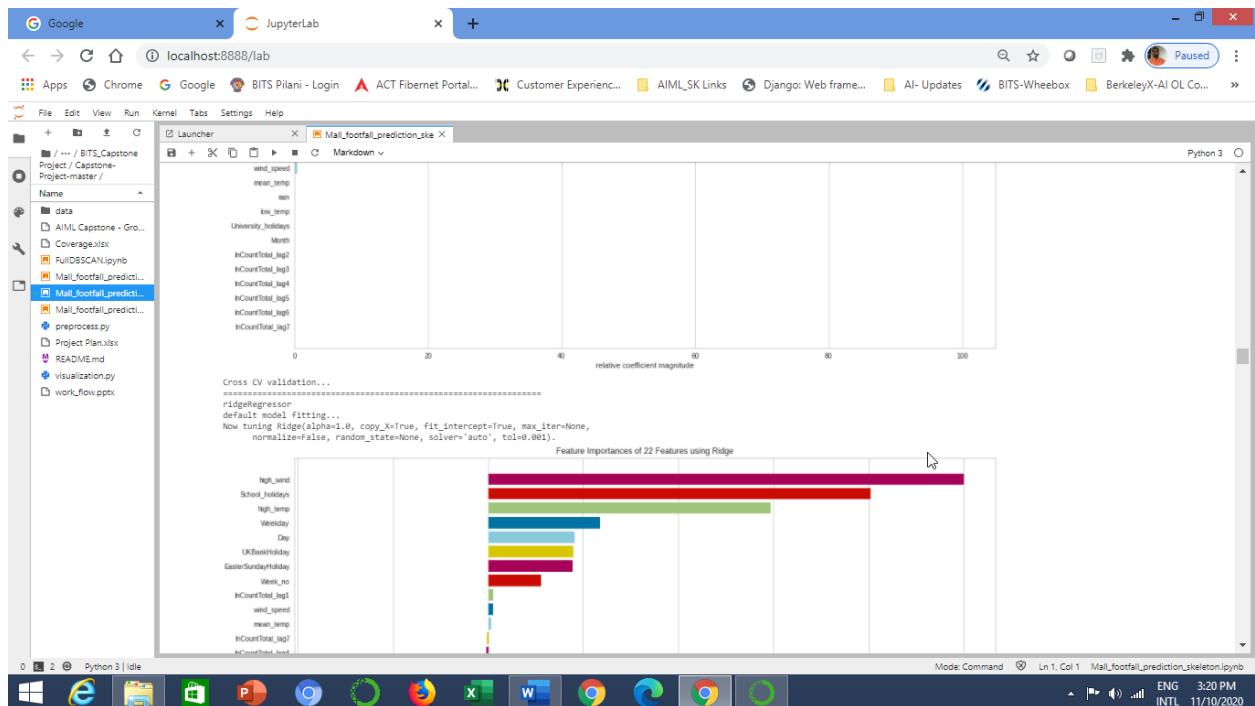
### 7.1.2 Model training with detrended data

```
[33]: reg_models = regression_model(data_final_detrended,dataset_type='detrend')
regression_models, modelPredictions, rscvPredictions, rfcvPredictions, loss_df_detrend = reg_models.train_and_evaluate_models(#cross_validation=False)

Defined 28 models
=====
linear_regression
default model fitting...
Now tuning LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False).
```

Feature Importances of 22 Features using LinearRegression





Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Python 3

```
wind_speed mean_temp
InCountTotal_lag1 InCountTotal_lag2
InCountTotal_lag3 InCountTotal_lag4
InCountTotal_lag5 InCountTotal_lag6
Month Jan
abnormal_rain inc_temp
University_holidays
```

Cross CV validation...

```
elastic_net
default model fitting...
Now tuning elasticNet(alpha=1.0, copy_X=True, fit_intercept=True, l1_ratio=0.5,
max_iter=100, normalize=False, positive=False, precompute=False,
random_state=None, selection='cyclic', tol=0.0001, warm_start=False).
```

Feature Importances of 22 Features using ElasticNet

Feature	Importance (approx.)
Ng_wend	95
Sched_holidays	65
Ng_temp	25
Weekday	15
Day	10
UKBankHoliday	8
EasterSundayHoliday	7
abnormal_rain	5
Week_no	5
InCountTotal_lag1	2
wind_speed	1

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 Python 3 | idle ENG 3:22 PM INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Python 3

```
wind_speed mean_temp
InCountTotal_lag1 InCountTotal_lag2
InCountTotal_lag3 InCountTotal_lag4
InCountTotal_lag5 InCountTotal_lag6
Month Jan
abnormal_rain inc_temp
University_holidays
```

Cross CV validation...

```
huberRegressor
default model fitting...
Now tuning HuberRegressor(alpha=0.0003, epsilon=1.35, fit_intercept=True, max_iter=100,
tol=1e-05, warm_start=False).
```

Feature Importances of 22 Features using HuberRegressor

Feature	Importance (approx.)
Weekday	95
Day	45
Sched_holidays	40
Week_no	30
EasterSundayHoliday	25
Ng_wend	10
UKBankHoliday	8
University_holidays	5
InCountTotal_lag1	5
Ng_temp	5
mean_temp	5
InCountTotal_lag2	2
InCountTotal_lag3	2
wind_speed	2
InCountTotal_lag4	1
inc_temp	1
InCountTotal_lag5	1
InCountTotal_lag6	1

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 Python 3 | idle ENG 3:22 PM INTL 11/10/2020

The screenshot shows a Jupyter Notebook interface with several tabs open. The active tab is titled "Mail\_footfall\_prediction\_skeleton.ipynb". The code cell contains the following Python code:

```
Cross CV validation...
passive_aggressor
default_model fitting...
Now tuning PassiveAggressiveRegressor(C=1.0, average=False, early_stopping=False,
epsilon=0.1, fit_intercept=True,
loss='epsilon_insensitive', max_iter=1000
n_iter_no_improve=5, random_state=None, shuffle=True,
tol=0.01, validation_fraction=0.1, verbose=0,
warm_start=False).
```

Below the code, a horizontal bar chart titled "Feature Importances of 22 Features using PassiveAggressiveRegressor" is displayed. The x-axis represents the relative coefficient magnitude, ranging from -20 to 100. The y-axis lists 22 features. The bars are color-coded by category: red for weekday, green for day, blue for week\_no, and yellow for other categories.

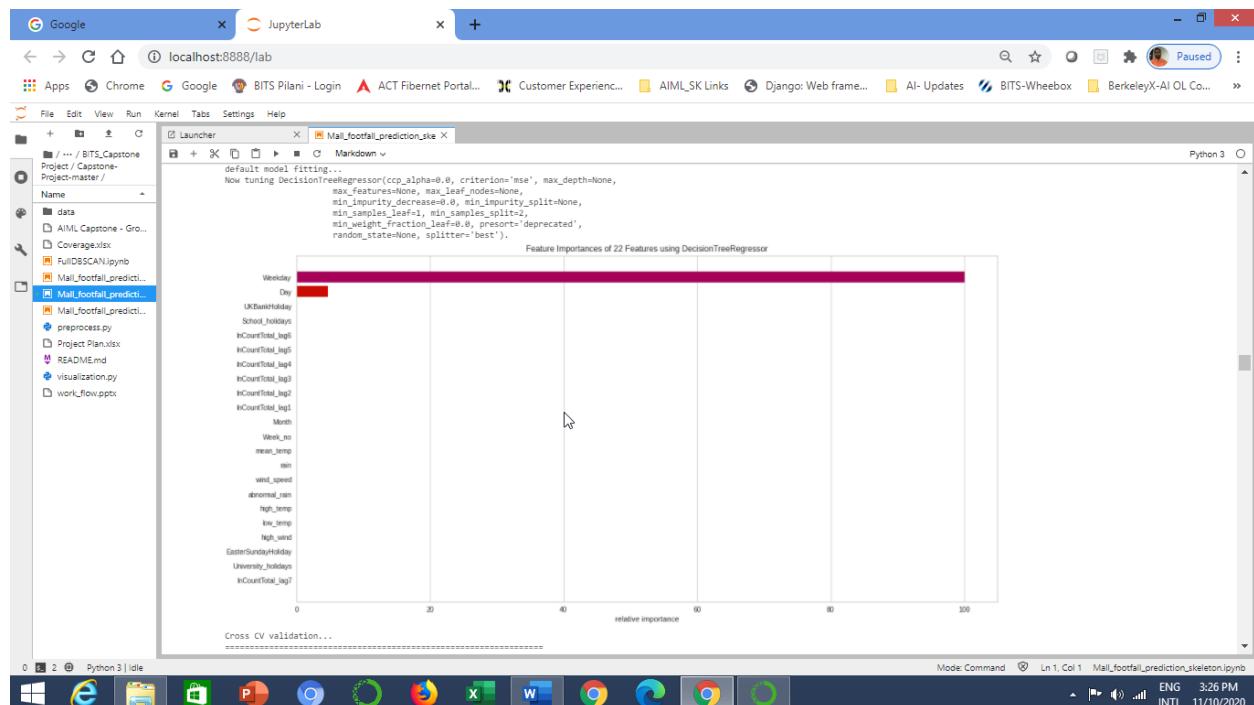
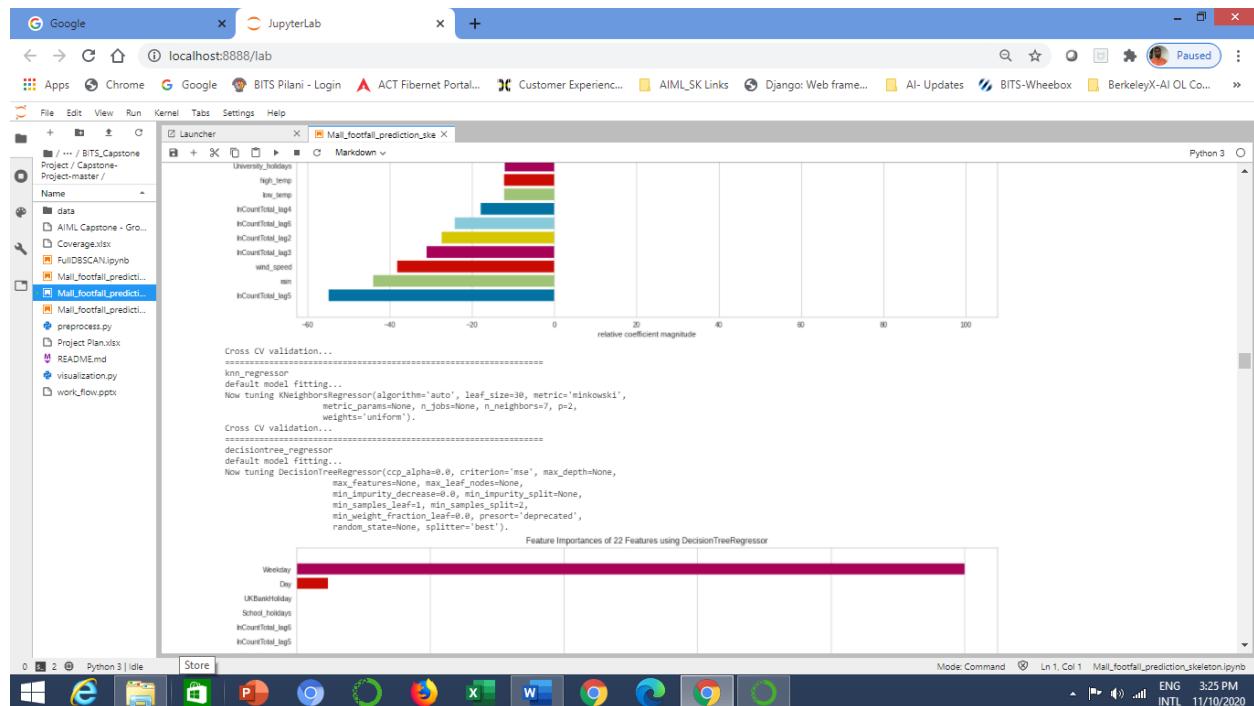
Feature	Importance Category
Weekday	Weekday
Day	Day
Week_no	Week_no
InCountTotal_jag1	InCountTotal_jag1
InCountTotal_jag1	InCountTotal_jag1
wind_speed	Other
EasterSundayHoliday	Other
mean_temp	Other
UKBankHoliday	Other
Month	Other
InCountTotal_jag1	InCountTotal_jag1
InCountTotal_jag3	InCountTotal_jag3
InCountTotal_jag4	InCountTotal_jag4
InCountTotal_jag2	InCountTotal_jag2
sun	Other
high_temp	Other
absenteeism_sun	Other
high_rain	Other
high_wind	Other
low_temp	Other
School_holidays	Other
University_holidays	Other
InCountTotal_jag1	InCountTotal_jag1

The screenshot shows a JupyterLab interface with a Python 3 kernel. The left sidebar displays a file tree for a 'BITS\_Capstone' project, including files like 'Coverage.xlsx', 'fullDBSCAN.ipynb', 'Mail\_footfall\_prediction.ipynb', 'preprocess.py', 'Project Plan.xlsx', 'README.md', and 'visualization.py'. The main area has two tabs open: 'Launcher' and 'Mail\_footfall\_prediction\_skeleton.ipynb'. The notebook tab contains code for cross-validation and SGDRegressor fitting, followed by a bar chart titled 'Feature Importances of 22 Features using SGDRegressor'. The x-axis is 'relative coefficient magnitude' ranging from -40 to 100. The y-axis lists features: InCountTotal\_lag7, Weekday, Day, InCountTotal\_lag1, Week\_no, meanTemp, Month, EasterSundayHoliday, UKBankerholiday, abnormal\_rain, School\_holiday, Nhg\_wind, University\_holiday, Nhg\_temp, kwh\_temp, InCountTotal\_lag1, InCountTotal\_lag2, and InCountTotal\_lag3. The most important feature is 'InCountTotal\_lag7' with a positive magnitude of approximately 95.

```
cross_val_score(..., cv=5)
SGDRegressor(alpha=0.0001, average=False, early_stopping=False, epsilon=0.1,
             eta0=0.01, fit_intercept=True, l1_ratio=0.15,
             learning_rate='invscaling', loss='squared_loss', max_iter=1000,
             n_iter_no_change=5, penalty='l2', power_t=0.25, random_state=None,
             shuffle=True, tol=0.001, validation_fraction=0.1, verbose=0,
             warm_start=False).
```

Feature Importances of 22 Features using SGDRegressor

Feature	Relative Coefficient Magnitude
InCountTotal_lag7	~95
Weekday	~35
Day	~25
InCountTotal_lag1	~20
Week_no	~15
meanTemp	~10
Month	~10
EasterSundayHoliday	~8
UKBankerholiday	~8
abnormal_rain	~8
School_holiday	~8
Nhg_wind	~8
University_holiday	~8
Nhg_temp	~8
kwh_temp	~8
InCountTotal_lag1	~5
InCountTotal_lag2	~5
InCountTotal_lag3	~5



The screenshot shows a JupyterLab interface with a Python 3 kernel. The code cell contains a snippet of code for tuning an ExtraTreeRegressor model:

```
extra_tree_regressor = ExtraTreeRegressor(cc_alpha=0.0, criterion='mse', max_depth=None,
                                         max_features='auto', max_leaf_nodes=None,
                                         min_impurity_decrease=0.0, min_impurity_split=None,
                                         min_samples_leaf=1, min_samples_split=2,
                                         min_weight_fraction_leaf=0.0, random_state=None,
                                         splitter='random').
```

Below the code is a chart titled "Feature Importances of 22 Features using ExtraTreeRegressor". The x-axis represents relative importance from 0 to 100. The y-axis lists features. The most important feature is "Weekday", with a value around 95. Other significant features include "Day" (~25), "Week\_no" (~15), and "School\_holidays" (~10). Many other features have very low importance values.

Feature	Relative Importance
Weekday	~95
Day	~25
Week_no	~15
School_holidays	~10
InCountTotal_1g1	~5
InCountTotal_1g2	~3
UKBankholiday	~2
Ngh_wend	~1
mean_temp	~1
rain	~1
InCountTotal_1g3	~1
InCountTotal_1g5	~1
InCountTotal_1g6	~1
abnormal_rain	~1
wind_speed	~1
ngh_temp	~1
low_temp	~1
Month	~1
EasterSunday/holiday	~1
University_holidays	~1
InCountTotal_1g7	~1
InCountTotal_1g8	~1

The screenshot shows a Jupyter Notebook interface with a sidebar containing project files and a main workspace displaying code and a bar chart.

**Code Output:**

```
Cross CV validation...
-----
svm_regress...  
default model fitting...  
Now tuning AdaBoostRegressor(base_estimator=None, learning_rate=1.0, loss='linear', n_estimators=200, random_state=None).  
Cross CV validation...
-----
adaBoost  
default model fitting ...  
Now tuning AdaBoostRegressor(base_estimator=None, learning_rate=1.0, loss='linear', n_estimators=200, random_state=None).
```

**Feature Importance Bar Chart:**

A horizontal bar chart titled "Feature Importances of 22 Features using AdaBoostRegressor". The x-axis represents the relative importance of each feature, ranging from 0.0 to 1.0. The y-axis lists the features. The bars are color-coded by category.

Feature	Relative Importance (approx.)
Day	0.95
Weekday	0.85
InCountTotal_1g5	0.75
Week_no	0.65
InCountTotal_1g3	0.55
InCountTotal_1g1	0.50
InCountTotal_1g3	0.45
InCountTotal_1g2	0.40
sun	0.35
UKBankHoliday	0.30
mean_temp	0.25
InCountTotal_1g7	0.20
wind_speed	0.18
InCountTotal_1g4	0.15
Month	0.12
EasterSundayHoliday	0.10
School_holidays	0.08
University_holidays	0.05
low_temp	0.03
high_wind	0.02
abnormal_rain	0.01
high_temp	0.01

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Project / Capstone-Project-master /

- Name
- data
- AIML Capstone - Gro...
- Coverage.xlsx
- fullDBSCAN.ipynb
- Mail\_footfall\_predict...
- Mail\_footfall\_predict...
- preprocess.py
- Project Plan.xlsx
- README.md
- visualization.py
- work\_flow.pptx

Feature Importances of 22 Features using AdaBoostRegressor

relative importance

```

Cross CV validation...
=====
bagging
default model fitting...
Now tuning BaggingRegressor(base_estimator=None, bootstrap=True, bootstrap_features=False,
max_features=1.0, max_samples=1.0, n_estimators=200,
n_jobs=None, oob_score=False, random_state=None, verbose=0,
warm_start=False).

Cross CV validation...

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3

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Launcher Mail\_footfall\_prediction\_ske

Project / Capstone-Project-master /

- Name
- data
- AIML Capstone - Gro...
- Coverage.xlsx
- fullDBSCAN.ipynb
- Mail\_footfall\_predict...
- Mail\_footfall\_predict...
- preprocess.py
- Project Plan.xlsx
- README.md
- visualization.py
- work\_flow.pptx

random\_forest

default model fitting...

Now tuning RandomForestRegressor(bootstrap=True, ccp\_alpha=0.0, criterion='mse',
max\_depth=None, max\_features='auto', max\_leaf\_nodes=None,
max\_samples=None, min\_impurity\_decrease=0.0,
min\_impurity\_split=None, min\_samples\_leaf=1,
min\_samples\_split=2, min\_weight\_fraction\_leaf=0.0,
n\_estimators=200, n\_jobs=None, oob\_score=False,
random\_state=None, verbose=0, warm\_start=False).

Feature Importances of 22 Features using RandomForestRegressor

relative importance

```

Cross CV validation...
=====
warm_start=False).

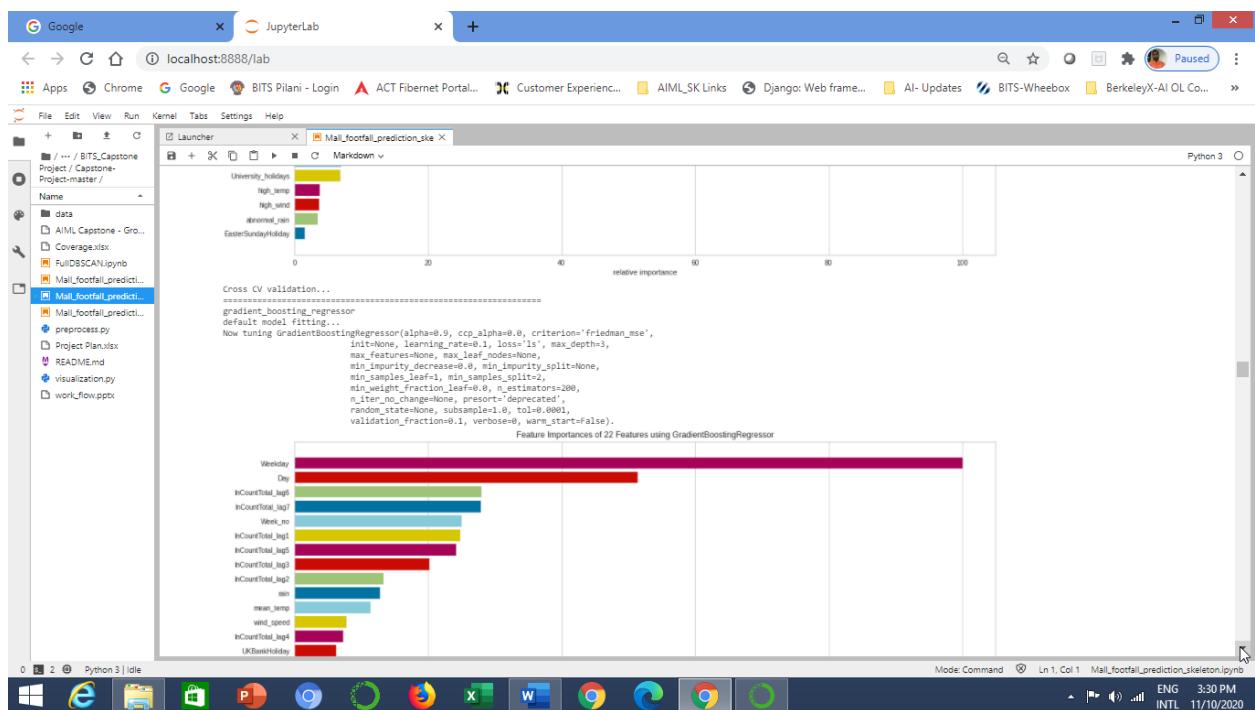
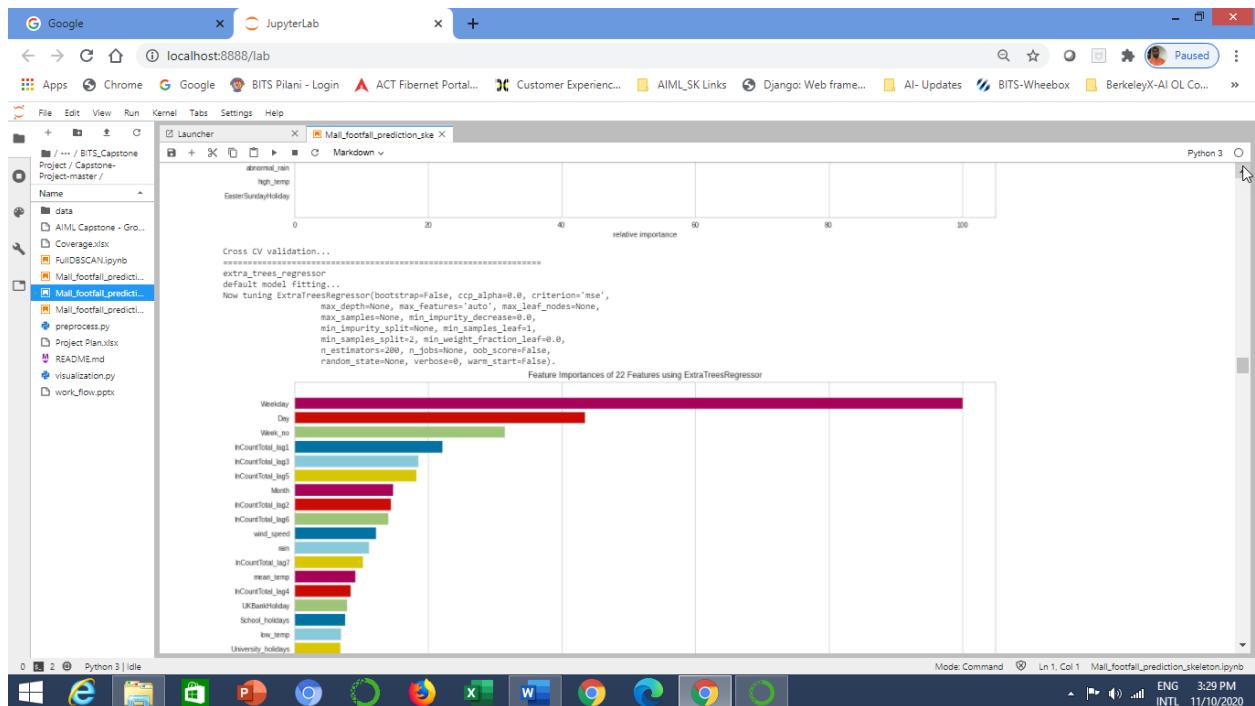
random_forest
default model fitting...
Now tuning RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
max_depth=None, max_features='auto', max_leaf_nodes=None,
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_leaf=0.0,
n_estimators=200, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False).

Cross CV validation...

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3



Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Project / Capstone-Project-master /

Name

data

AIML Capstone - Gro...

Coverage.xlsx

FullDBSCAN.ipynb

Mail\_footfall\_predict...

Mail\_footfall\_predict...

preprocess.py

Project Plan.xlsx

README.md

visualization.py

work\_flow.pptx

UKBankHoliday

Month

School\_holidays

University\_holidays

km\_temp

high\_wind

high\_temp

abnormal\_rain

EasterSundayHoliday

relative importance

Cross CV validation...

XGBoost

default model fitting...

```
Now tuning XGBRegressor(base_score=0.5, booster='gbtree', colsample_bytree=1,
colsample_bynode=1, colsample_bylevel=1, gamma=0,
importance_type='split', learning_rate=0.1, max_depth=6,
max_depth=3, min_child_weight=1, missing=None, n_estimators=100,
n_jobs=1, nthread=None, objective='reg:squarederror',
random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
seed=None, silent=None, subsample=1, verbosity=1).
```

Feature Importances of 22 Features using XGBRegressor

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

ENGLISH 3:31 PM INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Project / Capstone-Project-master /

Name

data

AIML Capstone - Gro...

Coverage.xlsx

FullDBSCAN.ipynb

Mail\_footfall\_predict...

Mail\_footfall\_predict...

preprocess.py

Project Plan.xlsx

README.md

visualization.py

work\_flow.pptx

in

InCountTotal\_leg1

UKBankHoliday

InCountTotal\_leg2

mean\_temp

abnormal\_rain

wind\_speed

School\_holidays

high\_temp

km\_temp

high\_wind

InCountTotal\_leg1

EasterSundayHoliday

University\_holidays

relative importance

Cross CV validation...

LightGBMBoosting

default model fitting...

```
Now tuning LGBMRegressor(booster_type='gbdt', class_weight=None, colsample_bytree=1.0,
importance_type='split', learning_rate=0.1, max_depth=-1,
min_child_weight=1, missing=None, n_estimators=100,
n_jobs=-1, num_leaves=31, objective='none',
random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
subsample=1.0, subsample_for_bin=200000, subsample_freq=0).
```

Cross CV validation...

CategoricalGradientBoosting

default model fitting...

```
Now tuning catboost.core.CatBoostRegressor object at 0x7f9271538d68.
```

Cross CV validation...

NeuralNet\_MLP

default model fitting...

```
Now tuning MLPRegressor(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
beta_2=0.99, early_stopping=False, epsilon=1e-08,
hidden_layer_sizes=[100], learning_rate='constant',
```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

ENGLISH 3:31 PM INTL 11/10/2020

The screenshot shows a JupyterLab interface running in a browser window. The top navigation bar includes tabs for 'Google', 'JupyterLab', and 'localhost:8888/lab'. Below the tabs, there are several browser tabs open, including 'BITS Pilani - Login', 'ACT Fibernet Portal...', 'Customer Experience...', 'AIIML\_SK Links', 'Django: Web frame...', 'AI - Updates', 'BITS-Wheebox', and 'BerkeleyX-AI OL Co...'. The main workspace contains two open notebooks:

- [75]:** A code cell containing the command `loss\_df\_detrend.sort\_values('mape').head(25)`.
- [75]:** The resulting output is a table with the following columns: Model\_type, mape, mse, me, mae, mpe, rmse, R2 score, AdjR2 score, AIC, and BIC. The table lists 34 different models, each with its specific parameters and performance metrics.

Model_type	mape	mse	me	mae	mpe	rmse	R2 score	AdjR2 score	AIC	BIC
extra_tree_regressor_detrend_default_mode	17502238	0.007274	0.005979	0.001004	0.042695	0.005272	0.194315	0.138607	-1711.69608	-1622.381943
CategoricalGradientBoosting_detrend_default_mode	24157379	0.004544	0.002549	0.004515	0.121964	0.007924	0.498615	0.462026	-1879.21992	-1790.634069
bagging_detrend_default_mode	21.831401	0.004567	0.002590	0.045331	0.127210	0.007577	0.494044	0.459098	-1877.861461	-1788.673538
random_forest_detrend_default_mode	21.865811	0.004483	0.003244	0.045142	0.159479	0.006952	0.503385	0.469062	-1884.499383	-1795.311461
CategoricalGradientBoosting_detrend_rscv_mode	25.199419	0.004584	0.002547	0.046457	0.153533	0.008121	0.485874	0.450386	-1872.142654	-1782.954731
extra_trees_regressor_detrend_rscv_mode	22.051522	0.004397	0.001248	0.048872	0.129843	0.0067124	0.500813	0.469335	-1891.380937	-1802.193014
extra_trees_regressor_detrend_default_mode	22.126706	0.004506	0.001522	0.044423	0.129843	0.0067124	0.500813	0.469335	-1882.670004	-1793.492081
gradient_boosting_regressor_detrend_rscv_mode	23.214041	0.004544	0.002049	0.044643	0.126512	0.008121	0.485643	0.450119	-1871.835935	-1782.795672
LightGradientBoosting_detrend_default_mode	23.165800	0.004491	0.000777	0.047814	0.138339	0.009933	0.493161	0.420736	-1833.399654	-1874.211931
LightGradientBoosting_detrend_rscv_mode	23.321795	0.004541	0.001695	0.046679	0.136005	0.007388	0.496870	0.462119	-1879.860799	-1790.672876
gradient_boosting_regressor_detrend_default_mode	23.199730	0.004820	0.019265	0.047341	0.136005	0.009428	0.495951	0.429095	-1883.599551	-1769.383228
random_forest_detrend_rscv_mode	23.533095	0.004797	0.000413	0.046653	0.136122	0.009257	0.489574	0.431860	-1860.27481	-1771.199538
ridgeRegressor_detrend_rscv_mode	23.740691	0.005505	0.001238	0.052431	0.121630	0.0074199	0.390033	0.347903	-1811.18549	-1721.920205
linear_regression_detrend_rscv_mode	23.747077	0.005506	0.001237	0.052433	0.121654	0.0074200	0.390012	0.347861	-1811.106144	-1721.918221
linear_regression_detrend_default_mode	23.747077	0.005506	0.001237	0.052433	0.121660	0.0074200	0.390012	0.347861	-1811.106144	-1721.918221
XGBoost_regressor_detrend_default_mode	24.225044	0.004897	0.001592	0.047183	0.145492	0.009975	0.457483	0.420012	-1852.953913	-1763.765590
huberRegressor_detrend_default_mode	24.266903	0.005684	0.004947	0.051830	0.118375	0.0072128	0.374659	0.331467	-1802.231743	-1713.048020
bagging_regressor_rscv_mode	24.340994	0.004896	0.000816	0.047891	0.146655	0.009975	0.457574	0.420102	-1833.039762	-1763.825749
XGBoost_regressor_rscv_mode	24.786844	0.005106	0.000765	0.050685	0.142728	0.0072012	0.423459	0.385776	-1832.790499	-1743.391126
huberRegressor_detrend_rscv_mode	25.307862	0.005795	0.004443	0.052337	0.130073	0.0076127	0.357915	0.313566	-1792.798514	-1703.610591
elastic_net_detrend_rscv_mode	25.372355	0.005441	0.000842	0.051494	0.142323	0.0073764	0.397163	0.335526	-1813.163030	-1726.124047
lassoRegressor_detrend_rscv_mode	25.399821	0.005440	0.000544	0.051468	0.142276	0.0073760	0.397232	0.335599	-1813.536827	-1726.168904
adaBoost_regressor_rscv_mode	26.222042	0.005159	0.001655	0.051097	0.182370	0.010782	0.428406	0.388926	-1834.314679	-1745.126755

The screenshot shows a JupyterLab environment with a terminal window open. The terminal window displays the following code and its execution results:

```
Cross CV validation...
=====
LightGBMGradientBoosting
default model fitting
Now tuning lgbm.Booster(boosting_type='gbdt', class_weight=None, colsample_bytree=1.0,
importance_type='split', learning_rate=0.1, max_depth=-1,
min_child_samples=20, min_child_weight=0.001, min_split_gain=0.0,
n_estimators=100, n_jobs=-1, num_leaves=31, objective=None,
random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
subsample=1.0, subsample_for_bin=200000, subsample_freq=0).

Cross CV validation...
=====
CategoricalGradientBoosting
default model fitting
Now tuning catboost.core.CatBoostRegressor object at 0x7f9271538d68>.
Cross CV validation...
=====
NeuralNet_MLP
default model fitting
Now tuning MLPRegression(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
beta_2=0.999, early_stopping=False, epsilon=1e-08,
hidden_layer_sizes=(100,), learning_rate='constant',
learning_rate_init=0.001, max_iter=200, max_n_ticks=200,
momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True,
power_t=0.5, random_state=None, shuffle=True, solver='adam',
tol=0.0001, validation_fraction=0.1, verbose=False,
warm_start=False).
Cross CV validation...
```

Google JupyterLab

localhost:8888/lab

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Launcher Mail\_footfall\_prediction\_ske

```
[34]: reg_models = regression_model(data_outliersremoved,dataset_type='Outliers_Removed')
      loss_values_all_models.append(loss_of_outlier_removed)
```

Defined 20 models

-----

Linear regression

Default model fitting...

Now tuning LinearRegression(copy\_X=True, fit\_intercept=True, n\_jobs=None, normalize=False).

Feature Importances of 22 Features using LinearRegression

Nhp\_wind  
School\_holidays  
Nhp\_temp  
Weekday  
University\_holidays  
Day  
Week\_no  
Month  
UKBankHoliday  
HCoutTotal\_lag1  
wind\_speed  
mean\_temp  
HCoutTotal\_lag2  
HCoutTotal\_lag3  
HCoutTotal\_lag4  
HCoutTotal\_lag5  
HCoutTotal\_lag6

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb ENG 3:34 PM INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

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Launcher Mail\_footfall\_prediction\_ske

```
[34]: reg_models = regression_model(data_outliersremoved,dataset_type='Outliers_Removed')
      loss_values_all_models.append(loss_of_outlier_removed)
```

Defined 20 models

-----

Lasso

Default model fitting...

Now tuning Lasso(copy\_X=True, fit\_intercept=True, max\_iter=1000, normalize=False, positive=False, precompute=False, random\_state=None, selection='cyclic', tol=0.0001, warm\_start=False).

Feature Importances of 22 Features using Lasso

HCoutTotal\_lag1  
HCoutTotal\_lag2  
HCoutTotal\_lag3  
HCoutTotal\_lag4  
HCoutTotal\_lag5  
HCoutTotal\_lag6  
EasterSundayHoliday  
abnormal\_Jan  
low\_temp

Cross CV validation...

-----

LassoRegressor

Default model fitting...

Now tuning Lasso(copy\_X=True, fit\_intercept=True, max\_iter=1000, normalize=False, positive=False, precompute=False, random\_state=None, selection='cyclic', tol=0.0001, warm\_start=False).

Feature Importances of 22 Features using Lasso

Nhp\_wind  
School\_holidays  
Nhp\_temp  
Weekday  
Day  
abnormal\_Jan  
Week\_no  
Month  
University\_holidays  
HCoutTotal\_lag1  
UKBankHoliday  
HCoutTotal\_lag2  
HCoutTotal\_lag3  
HCoutTotal\_lag4  
HCoutTotal\_lag5  
HCoutTotal\_lag6  
min

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb ENG 3:45 PM INTL 11/10/2020

Google JupyterLab

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File Edit View Run Kernel Tabs Settings Help

Launcher Markdown Python 3

```

mean_temp
InCountTotal_leg1
InCountTotal_leg2
ian
km_temp
EasterSundayHoliday
InCountTotal_leg3
InCountTotal_leg4

```

Cross CV validation...

```

ridgeRegresso
default model fitting...
Now tuning Ridge(alpha=1.0, copy_X=True, fit_intercept=True, max_iter=None,
normalize=False, random_state=None, solver='auto', tol=0.001).

```

Feature Importances of 22 Features using Ridge

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 0 Python 3 | idle ENG 3:46 PM INTL 11/10/2020

Google JupyterLab

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File Edit View Run Kernel Tabs Settings Help

Launcher Markdown Python 3

```

InCountTotal_leg5
InCountTotal_leg6
InCountTotal_leg7
ian
EasterSundayHoliday
km_temp
abnormal_rain

```

Cross CV validation...

```

elasticNet(alpha=1.0, copy_X=True, fit_intercept=True, l1_ratio=0.5,
max_iter=1000, normalize=False, positive=False, precompute=False,
random_state=None, selection='cyclic', tol=0.0001, warm_start=False).

```

Feature Importances of 22 Features using ElasticNet

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 0 Python 3 | idle ENG 3:47 PM INTL 11/10/2020

The screenshot shows a JupyterLab interface with a sidebar containing project files and a main workspace. The workspace displays a histogram titled 'Feature Importances of 22 Features using HuberRegressor'. The x-axis represents the relative coefficient magnitude, ranging from 0 to 100. The y-axis lists features: Weekday, Week\_no, Day, School\_holidays, Month, high\_wind, InCountTotal\_1st, University\_holidays, InCountTotal\_2nd, high\_temp, InCountTotal\_3rd, InCountTotal\_4th, InCountTotal\_5th, mean\_temp, and InCountTotal\_6th. The bars show that 'Weekday' has the highest importance (~95), followed by 'Day' (~55), 'Month' (~45), and 'high\_temp' (~35).

The screenshot shows a JupyterLab interface with a Python 3 kernel. The left sidebar displays a file tree for a project named 'BITS\_Capstone'. The main area contains a 'Launcher' tab showing a bar chart titled 'Feature Importances of 22 Features using PassiveAggressiveRegressor'. The x-axis is 'relative coefficient magnitude' ranging from -20 to 100. The y-axis lists features: 'IsCountTotal\_isAp1', 'mean\_temp', 'IsCountTotal\_isAp2', 'low\_temp', 'wind\_speed', 'UKBankHoliday', 'min', 'abnormal\_rain', 'EasterSundayHoliday'. The bars show that 'EasterSundayHoliday' has the highest positive magnitude (~90), while 'IsCountTotal\_isAp1' has the highest negative magnitude (~-18). Below the chart is a code cell for 'Cross CV validation...' using a PassiveAggressiveRegressor with C=1.0, average=False, and early\_stopping=True.

```
Cross CV validation...
=====
passive_aggressor
default model fitting...
Now tuning PassiveAggressiveRegressor(C=1.0, average=False, early_stopping=False,
epsilon=0.1, fit_intercept=True,
loss='epsilon_insensitive', max_iter=10000,
n_iter_no_change=5, random_state=None, shuffle=True,
tol=0.001, validation_fraction=0.1, verbose=0,
warm_start=False).
```

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Python 3

relative coefficient magnitude

Cross CV validation...

```
sgd
default model fitting...
Now tuning SGDRegressor(alpha=0.001, average=False, early_stopping=False, epsilon=0.1,
eta0=0.001, fit_intercept=True, l1_ratio=0.15,
learning_rate='invscaling', loss='squared_loss', max_iter=1000,
n_iter_no_change=5, penalty='l2', power_t=0.25, random_state=None,
shuffle=True, tol=0.001, validation_fraction=0.1, verbose=0,
warm_start=False).
```

Feature Importances of 22 Features using SGDRegressor

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

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Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Python 3

relative coefficient magnitude

Cross CV validation...

```
knn_regressor
default model fitting...
Now tuning KNeighborsRegressor(algorithm='auto', leaf_size=30, metric='minkowski',
metric_params=None, n_jobs=None, n_neighbors=7, p=2,
weights='uniform').
```

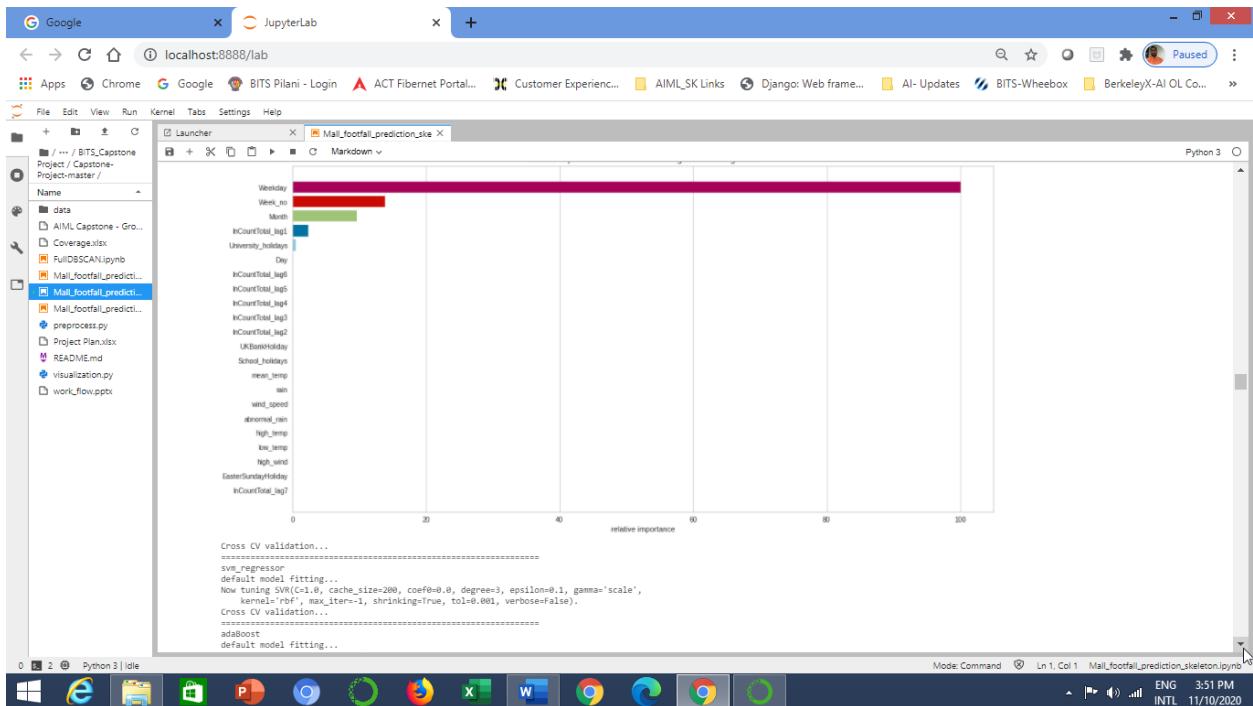
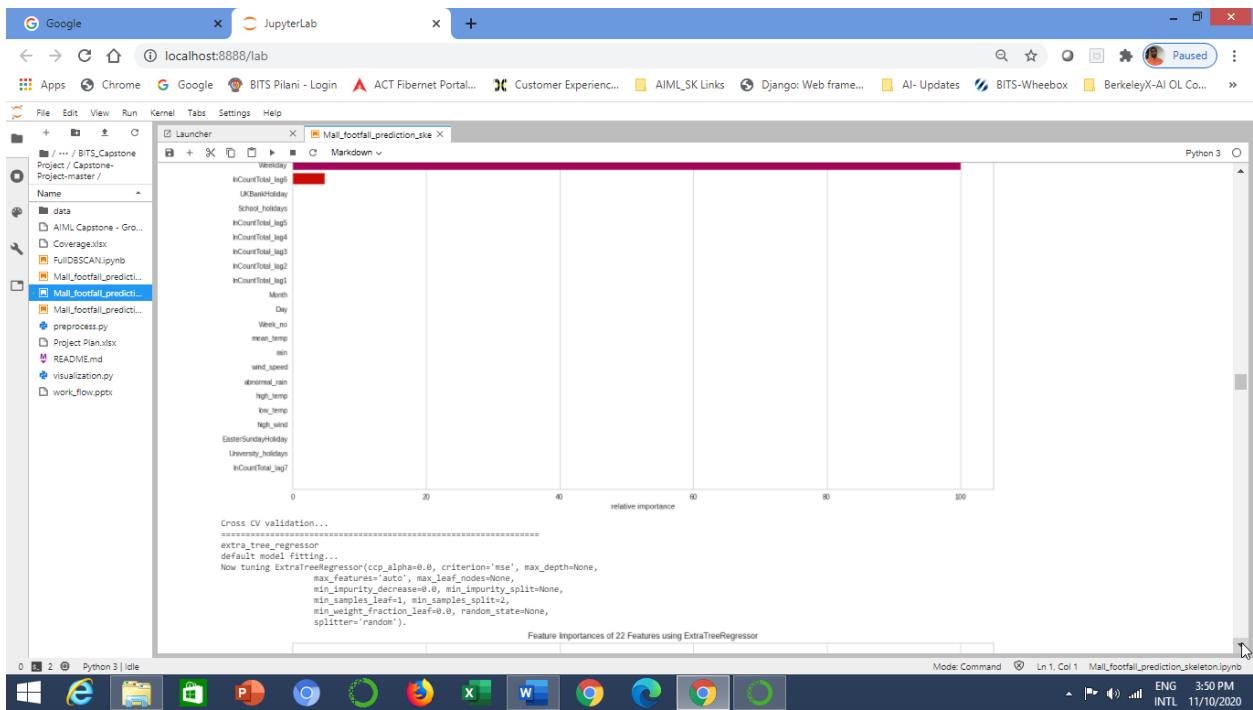
Cross CV validation...

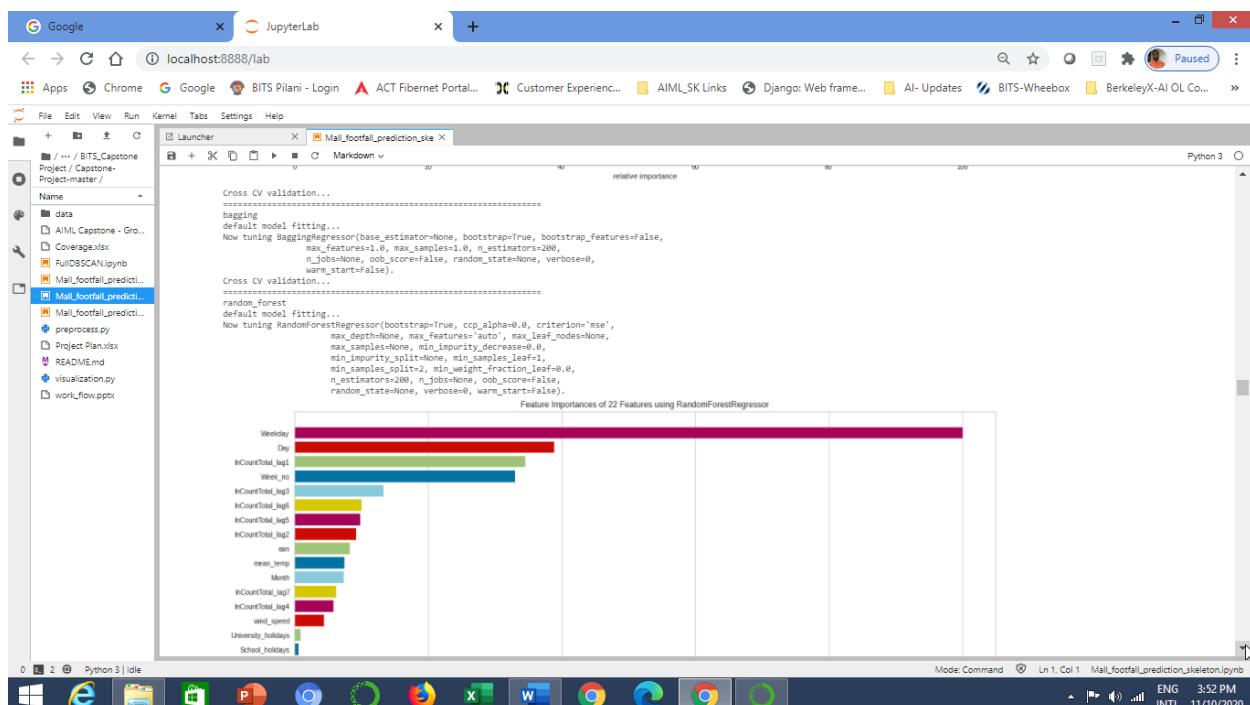
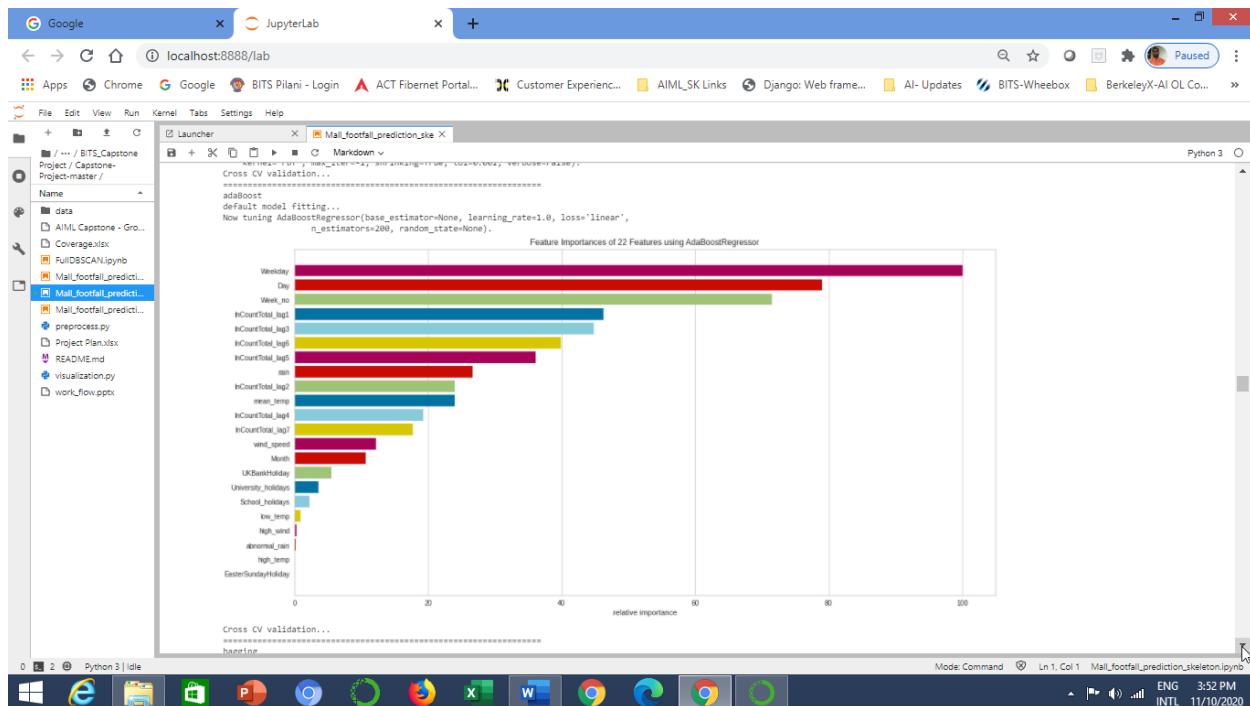
```
decisiontree_regressor
default model fitting...
Now tuning DecisionTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, presort='deprecated',
random_state=None, splitter='best').
```

Feature Importances of 22 Features using DecisionTreeRegressor

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

ENGLISH 3:50 PM INTL 11/10/2020





Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_skeleton

```

School_holiday
hgt_temp
low_temp
high_wind
abnormal_rain
EasterSundayHoliday
UKBankHoliday

```

relative importance

Cross CV validation...

```

extra_trees_regressor
default model fitting...
Now tuning ExtraTreesRegressor(ccp_alpha=0.0, criterion='mse',
                                max_depth=None, max_features='auto',
                                max_leaf_nodes=None,
                                max_depth=None, max_features='auto',
                                min_impurity_decrease=0.0,
                                min_impurity_split=None, min_samples_leaf=1,
                                min_samples_split=2, min_weight_fraction_leaf=0.0,
                                n_estimators=200, n_jobs=None, oob_score=False,
                                random_state=None, verbose=0, warm_start=False),

```

Feature Importances of 22 Features using ExtraTreesRegressor

Feature	Relative Importance
Weekday	~100
Week_no	~100
Day	~30
HCountTotal_lag1	~20
HCountTotal_lag2	~18
Month	~15
HCountTotal_lag3	~15
HCountTotal_lag4	~15
wind_speed	~12
HCountTotal_lag5	~12
rain	~10
HCountTotal_lag6	~10
mean_temp	~10
HCountTotal_lag7	~10
University_holiday	~5

Python 3

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

ENGLISH 3:53 PM INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_skeleton

```

University_holiday
HCountTotal_lag1
low_temp
School_holiday
high_wind
hgt_temp
abnormal_rain
UKBankHoliday
EasterSundayHoliday

```

relative importance

Cross CV validation...

```

gradient_boosting_regressor
default model fitting...
Now tuning GradientBoostingRegressor(alpha=0.9, ccp_alpha=0.0, criterion='friedman_mse',
                                         int=None, learning_rate=0.1, loss='ls', max_depth=3,
                                         max_features=None, max_leaf_nodes=None,
                                         min_impurity_decrease=0.0, min_impurity_split=None,
                                         min_leaf_node_size=None, min_weight_fraction_leaf=0.0,
                                         n_estimators=200, n_iter_no_change=None, presort='deprecated',
                                         random_state=None, subsample=1.0, tol=0.0001,
                                         validation_fraction=0.1, verbose=0, warm_start=False),

```

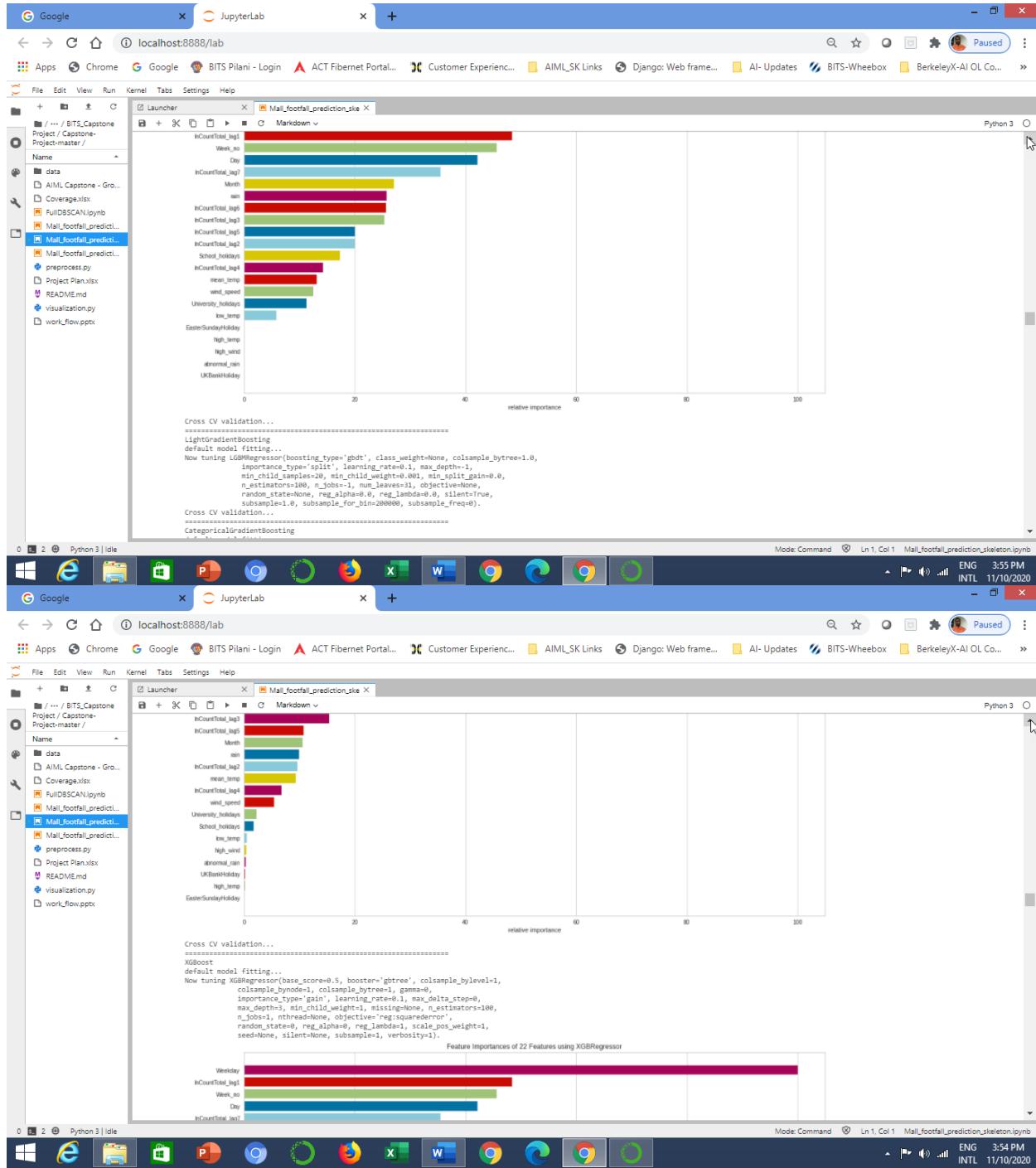
Feature Importances of 22 Features using GradientBoostingRegressor

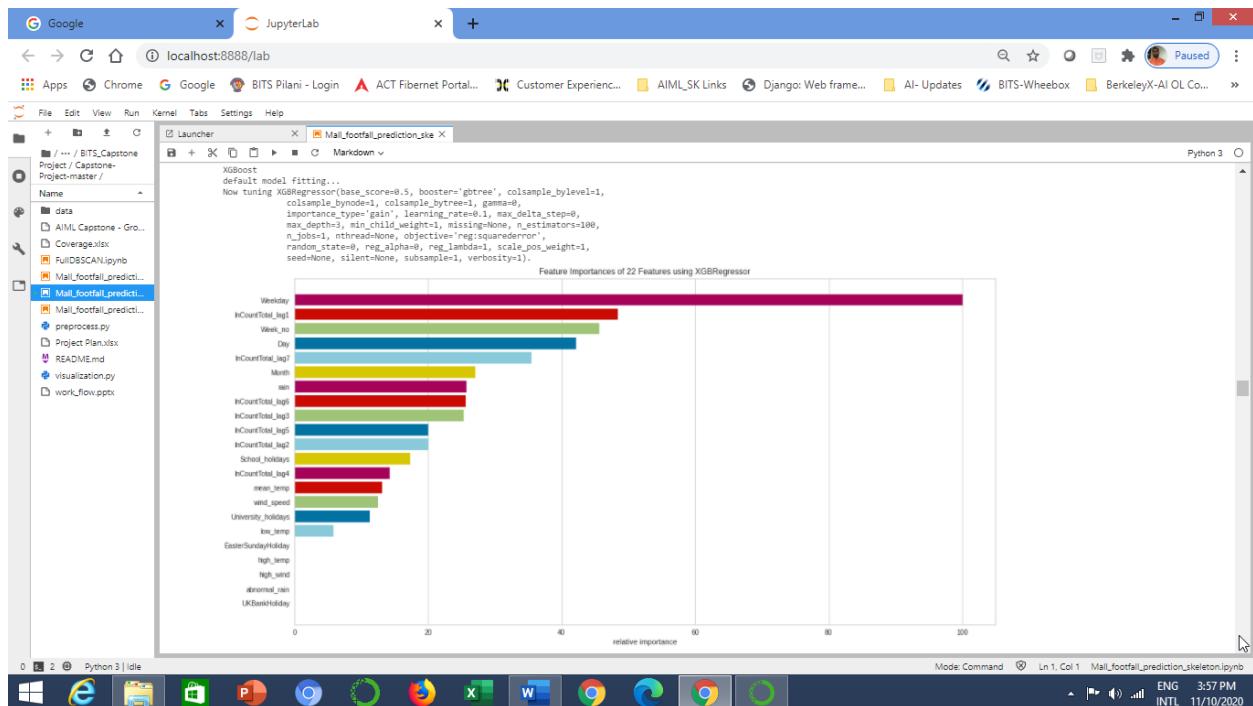
Feature	Relative Importance
Weekday	~100
Day	~30
Week_no	~25
HCountTotal_lag1	~20
HCountTotal_lag2	~18
HCountTotal_lag3	~18
HCountTotal_lag4	~15
HCountTotal_lag5	~15
Month	~12

Python 3

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

ENGLISH 3:54 PM INTL 11/10/2020





The screenshot shows a JupyterLab interface with a Python 3 kernel. The code cell contains cross-validation code for LightGBM and CatBoost models.

```

Cross CV validation...
=====
LightGBMBoosting
default model fitting...
Now tuning LightGBMBoosting(base_tuning_type='gbdt', class_weight=None, colsample_bytree=1.0,
importance_type='split', learning_rate=0.1, max_depth=-1,
min_child_samples=20, min_child_weight=0.001, min_split_gain=0.0,
n_estimators=100, n_jobs=-1, num_leaves=31, objective=None,
random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
subsample=1.0, subsample_for_bin=2000000, subsample_freq=0).

Cross CV validation...
=====
CategoricalGradientBoosting
default model fitting...
Now tuning CategoricalGradientBoosting object at @x7f92714d76d8.
Cross CV validation...
=====

NeuralNet_MLP
default model fitting...
Now tuning NeuralNet_MLP(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
beta_2=0.999, early_stopping=False, epsilon=1e-08,
hidden_layer_sizes=(100,), learning_rate='constant',
learning_rate_init=0.001, max_fun=15000, max_iter=200,
momentum=0.9, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam',
tol=0.0001, validation_fraction=0.1, verbose=False,
warm_start=False).

Cross CV validation...

```

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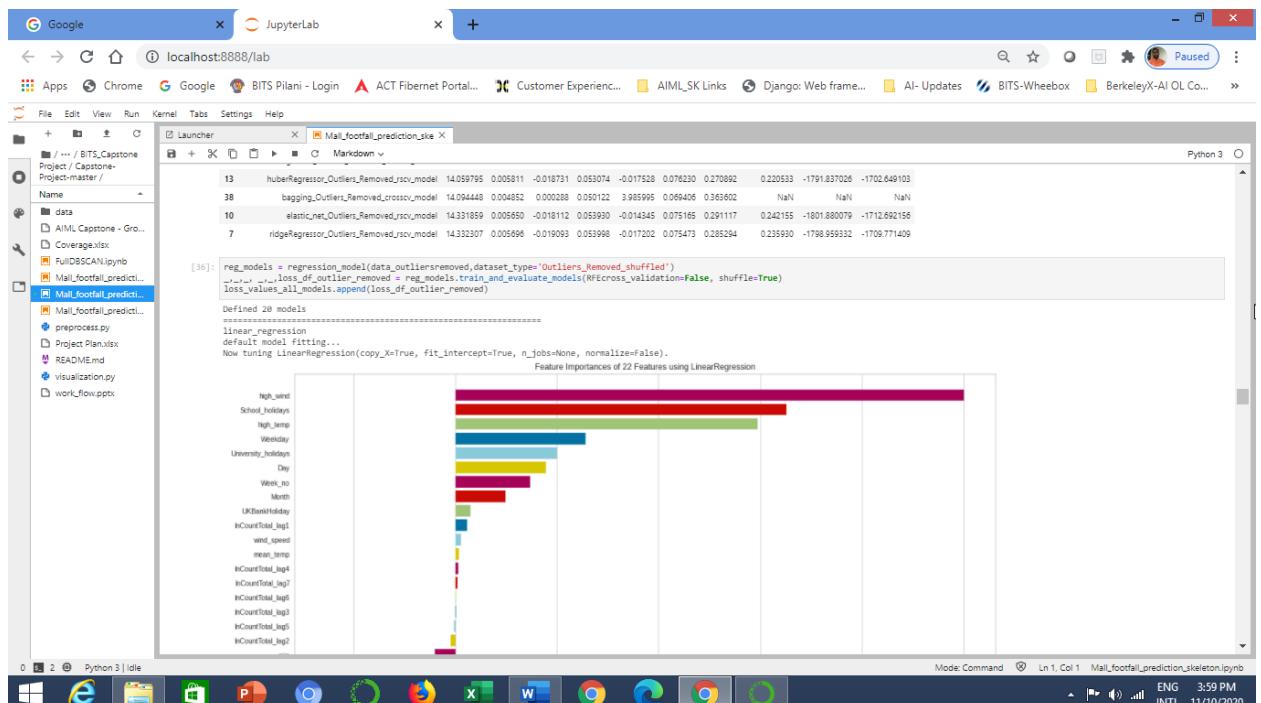
localhost:8888/lab

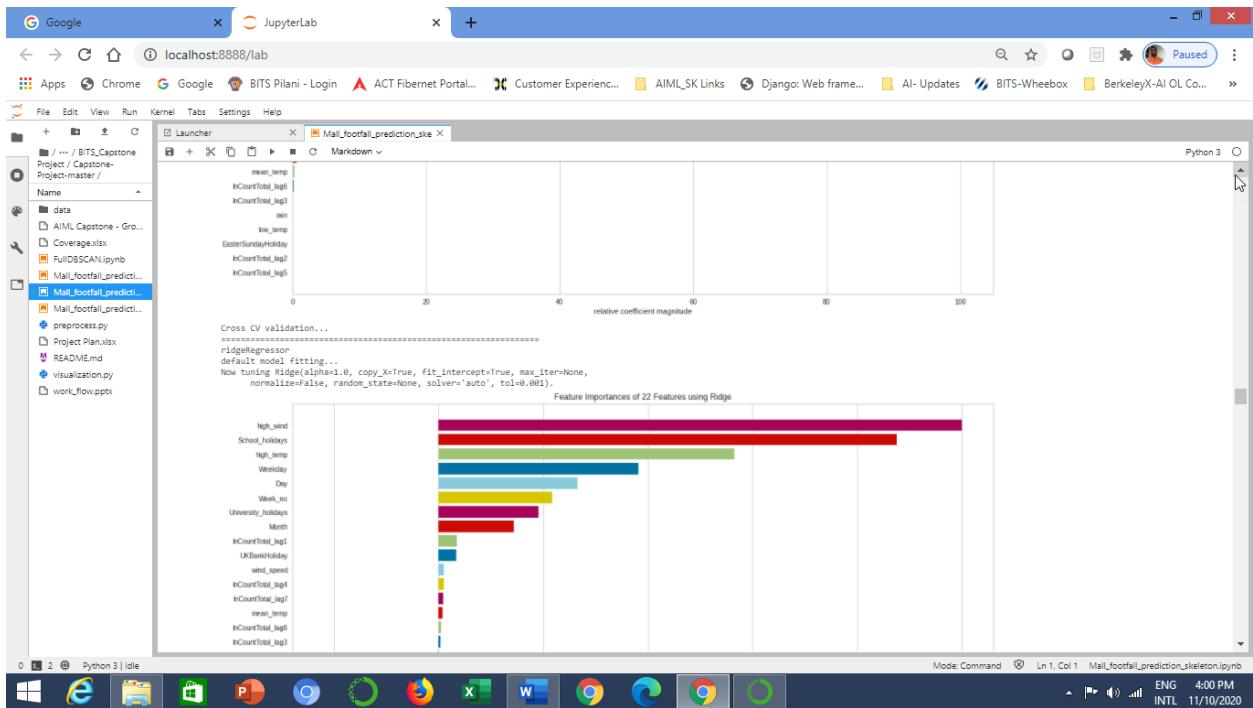
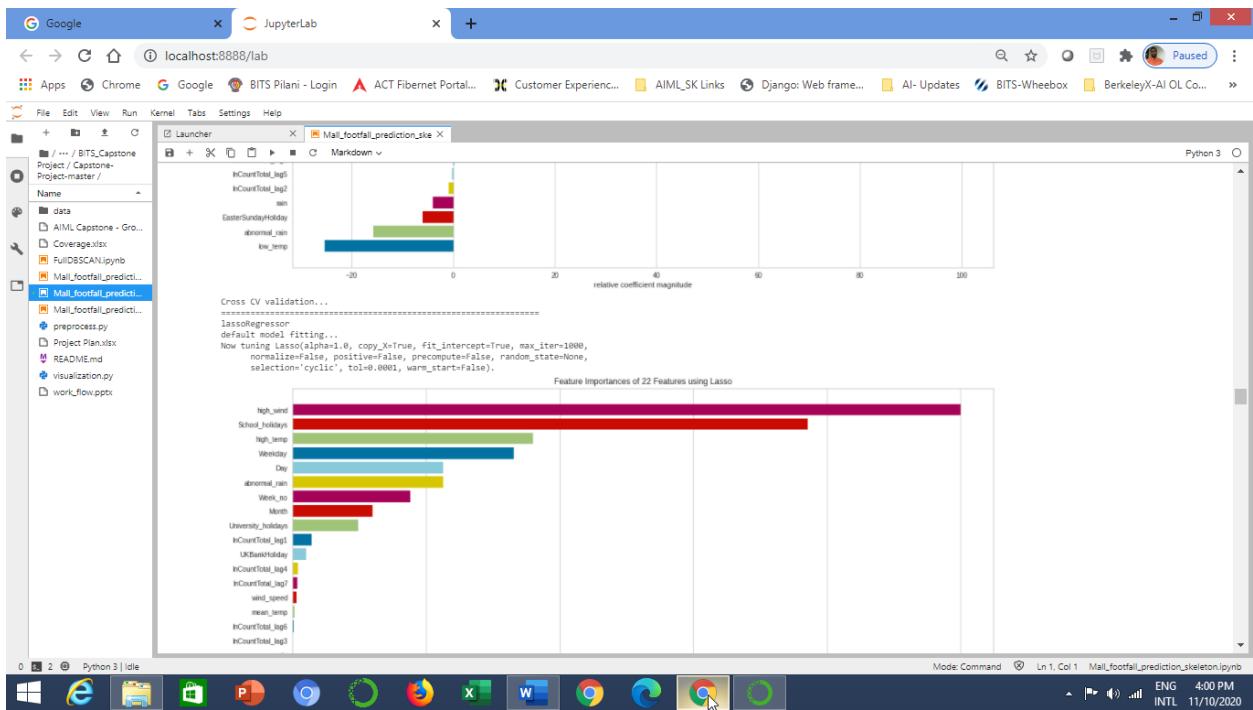
File Edit View Run Kernel Tabs Settings Help

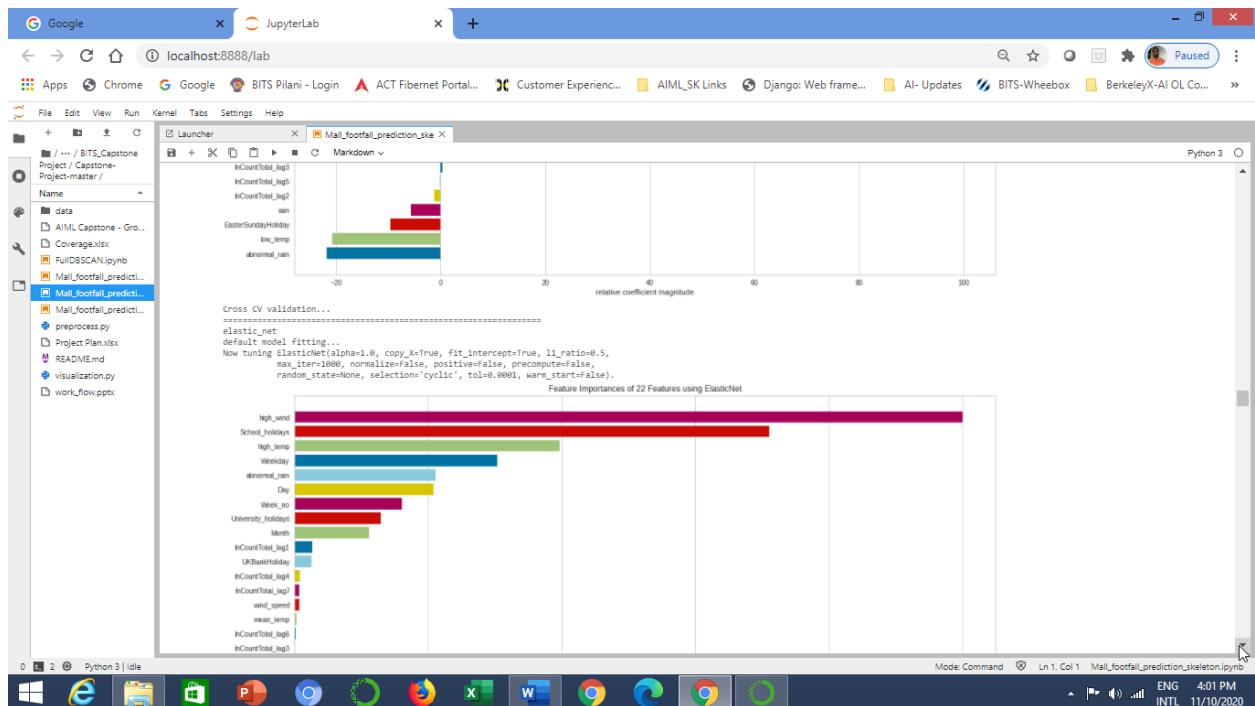
Launcher Mail\_footfall\_prediction\_ske

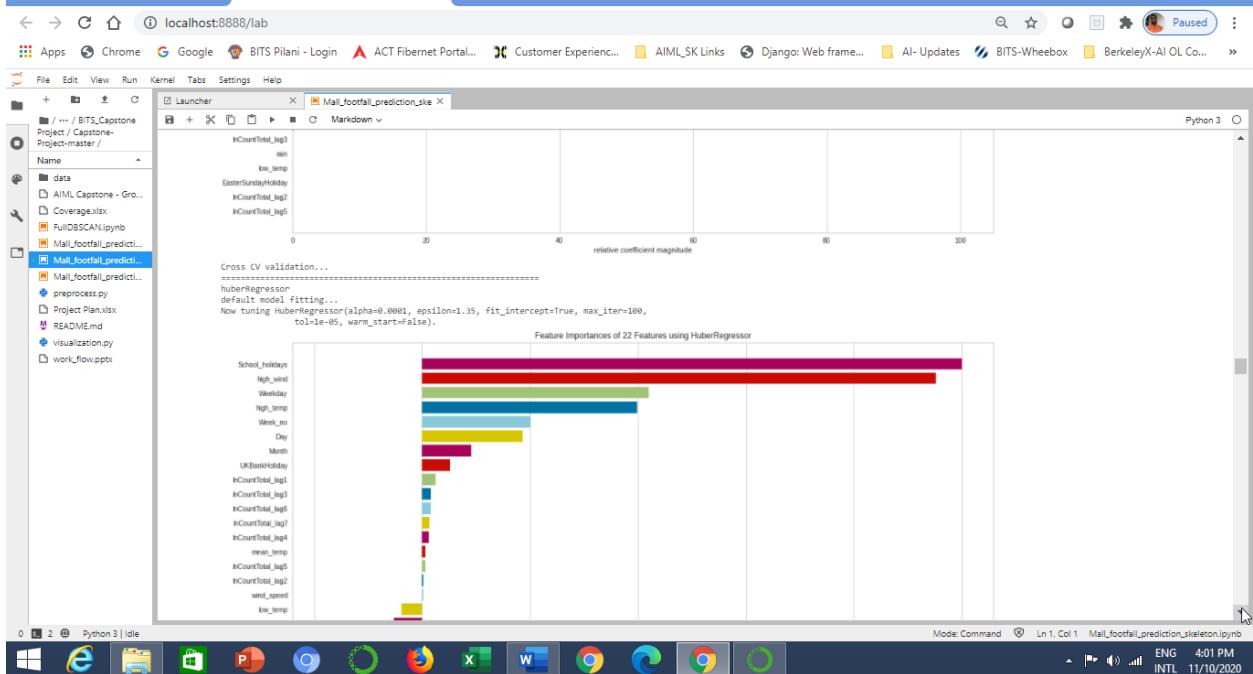
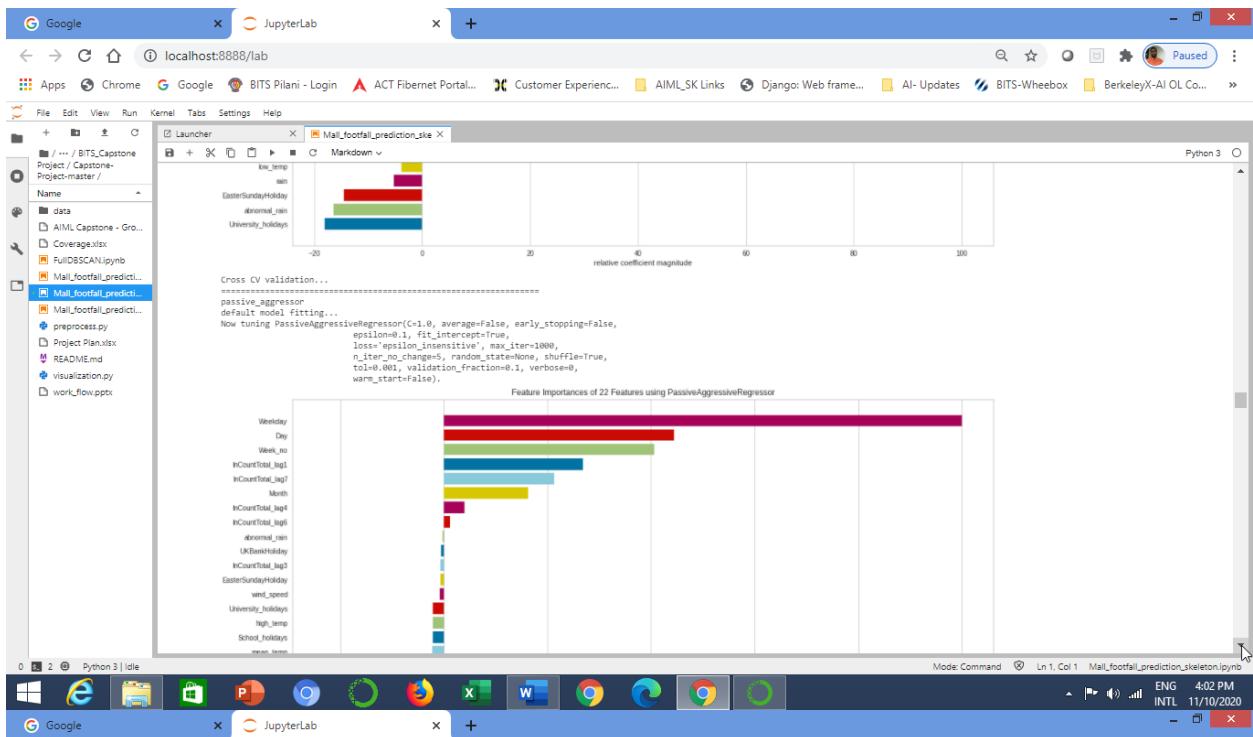
```
[35]: loss_df_outlier_removed.sort_values('mape').head(25)
```

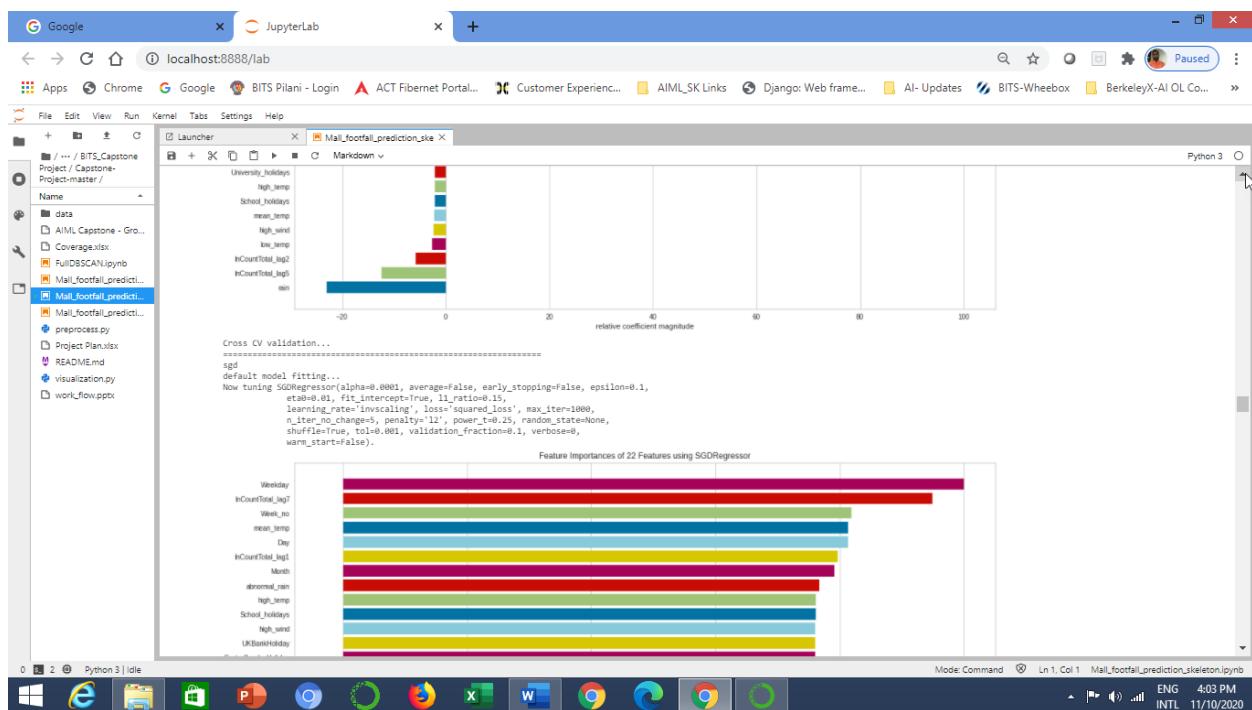
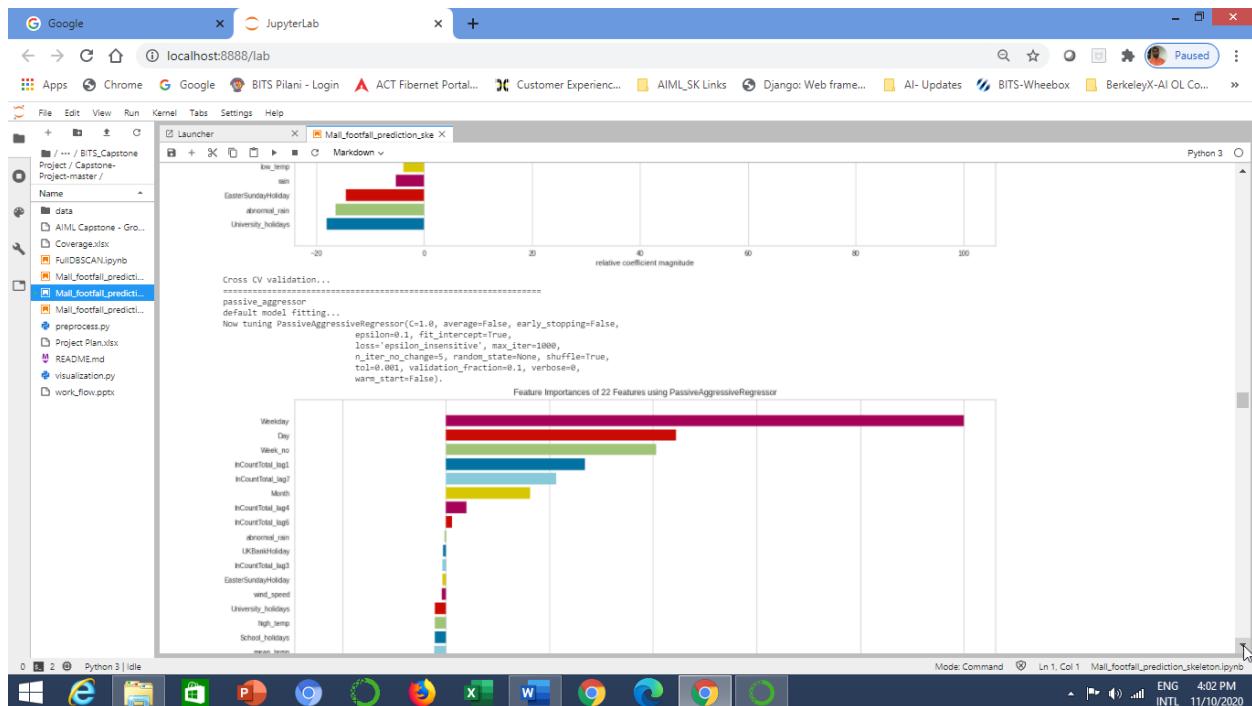
	Model_type	mape	mse	me	mae	mpe	r2 score	Adj R2 score	AIC	BIC
36	bagging_Outliers_Removed_default_model	11.815070	0.004205	-0.013330	0.044794	-0.009810	0.054849	0.472342	0.435897	-1907.581280
39	random_forest_Outliers_Removed_default_model	11.875625	0.004131	-0.012771	0.044793	-0.007844	0.064274	0.481667	0.445866	-1913.646660
54	CategoricalGradientBoosting_Outliers_Removed_c...	11.991476	0.004348	-0.012474	0.045395	-0.007204	0.059597	0.454849	0.416815	-1895.408784
40	random_forest_Outliers_Removed_rsmo_model	12.020205	0.004409	-0.015048	0.045884	-0.021278	0.066420	0.446783	0.408573	-1890.394464
45	gradient_boosting_regressor_Outliers_Removed_d...	12.205760	0.004710	-0.017632	0.047270	-0.021842	0.058532	0.408963	0.381654	-1866.800007
52	LightGradientBoosting_Outliers_Removed_rsmo_model	12.205936	0.004421	-0.016398	0.046592	-0.017115	0.066490	0.444593	0.406991	-1889.440826
42	extra_trees_regressor_Outliers_Removed_default...	12.529232	0.004405	-0.015248	0.047465	-0.014811	0.065345	0.447709	0.409563	-1890.920557
43	extra_trees_regressor_Outliers_Removed_rsmo_model	12.538173	0.004435	-0.015747	0.047570	-0.015819	0.066545	0.443570	0.405188	-1888.232707
48	XGBoost_Outliers_Removed_default_model	12.577377	0.004815	-0.017862	0.048142	-0.018978	0.069287	0.395919	0.354176	-1858.993631
46	gradient_boosting_regressor_Outliers_Removed_d...	12.674935	0.004782	-0.017924	0.048970	-0.018652	0.069151	0.400024	0.338894	-1861.427996
55	CategoricalGradientBoosting_Outliers_Removed_d...	12.691450	0.004485	-0.018857	0.047479	-0.008073	0.066570	0.443965	0.405561	-1888.830893
51	LightGradientBoosting_Outliers_Removed_default...	12.709911	0.004599	-0.014983	0.048137	-0.014091	0.067815	0.422397	0.383131	-1875.352644
44	extra_trees_regressor_Outliers_Removed_crossov...	12.759066	0.004284	-0.016384	0.045323	3.553093	0.065373	0.424719	NaN	NaN
37	bagging_Outliers_Removed_rsmo_model	13.075584	0.004868	-0.016489	0.048959	-0.009537	0.069791	0.389914	0.347776	-1855.402495
34	adaBoost_Outliers_Removed_crossov_model	13.114304	0.004669	-0.010281	0.048754	0.005294	0.068328	0.414209	0.373749	-1869.598628
49	XGBoost_Outliers_Removed_rsmo_model	13.225600	0.004730	-0.010576	0.049082	0.005688	0.068771	0.406566	0.365599	-1865.535668
41	random_forest_Outliers_Removed_crossov_model	13.306124	0.004550	-0.011445	0.047095	3.838474	0.057111	0.399523	NaN	NaN
47	gradient_boosting_regressor_Outliers_Removed_d...	13.352104	0.004584	-0.000497	0.047876	3.300756	0.067519	0.384451	NaN	NaN
56	CategoricalGradientBoosting_Outliers_Removed_d...	13.626440	0.004750	0.000092	0.048497	3.750387	0.066608	0.378603	NaN	NaN
35	adaBoost_Outliers_Removed_crossov_model	13.835259	0.004794	-0.001173	0.049539	4.010600	0.068853	0.373306	NaN	NaN
12	huberRegressor_Outliers_Removed_crossov_model	13.974179	0.005783	-0.018115	0.052393	-0.016486	0.075820	0.278586	0.228733	-1795.612351
13	huberRegressor_Outliers_Removed_rsmo_model	14.059795	0.005811	-0.018731	0.053074	-0.017528	0.076230	0.270892	0.220533	-1791.837026
38	bagging_Outliers_Removed_crossov_model	14.094448	0.004852	0.000288	0.050122	3.985995	0.069406	0.365602	NaN	NaN
10	elastic_net_Outliers_Removed_rsmo_model	14.331850	0.005650	-0.018112	0.059390	-0.014345	0.075165	0.291117	0.242155	-1801.880079
7	ridgeRegressor_Outliers_Removed_rsmo_model	14.332307	0.005696	-0.019093	0.053998	-0.017202	0.075473	0.285294	0.235930	-1798.959332

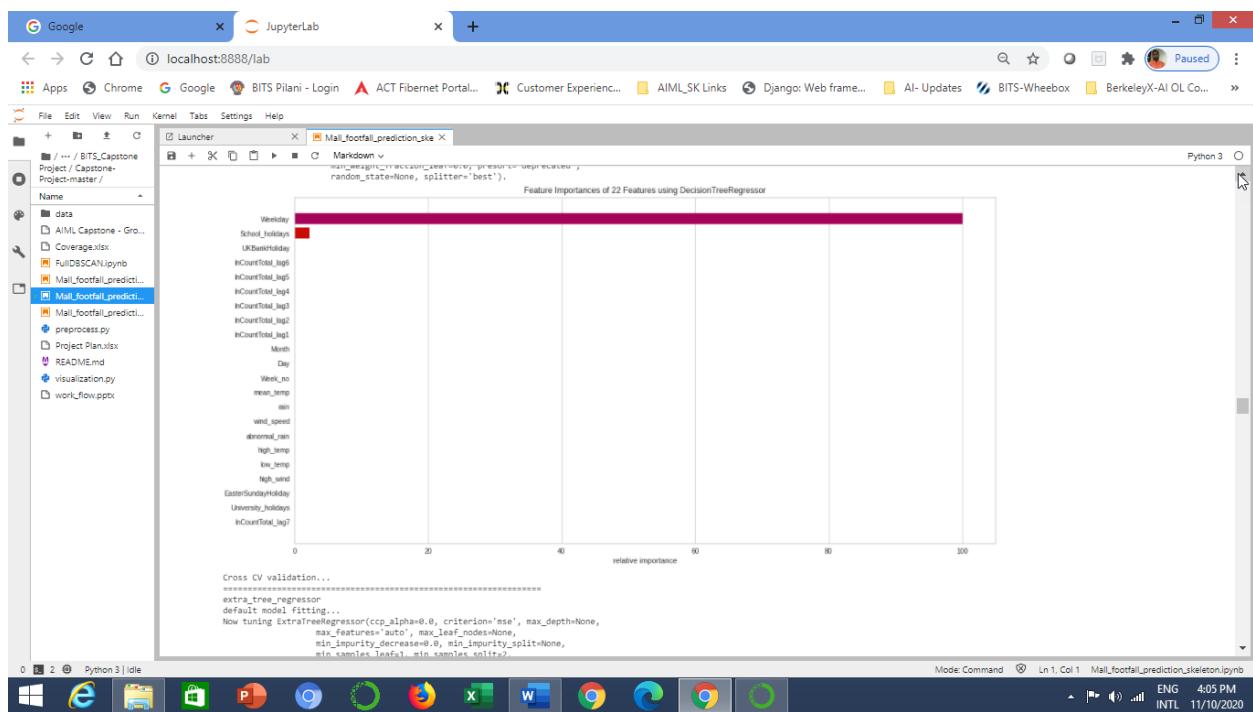
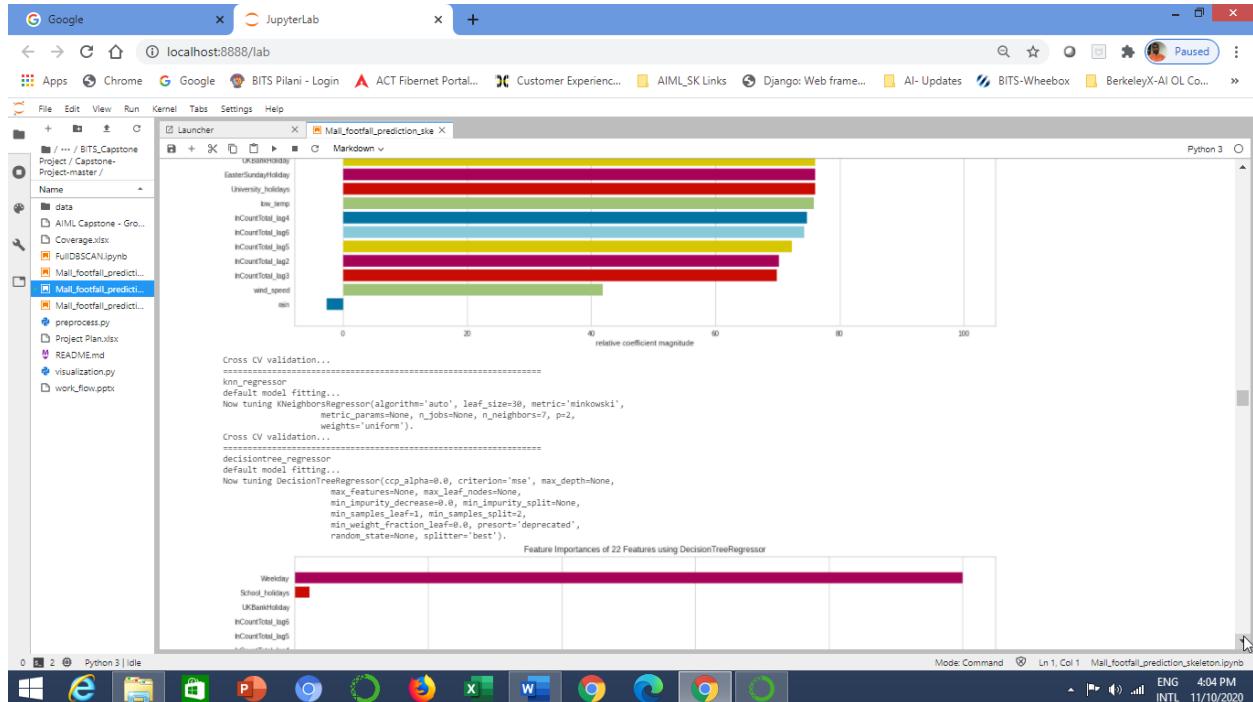












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Launcher Mail\_footfall\_prediction\_skeleton.py

```
default model fitting...
Now tuning ExtraTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,
max_features='auto', max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, random_state=None,
splitter='random').
```

Feature Importances of 22 Features using ExtraTreeRegressor

Cross CV validation...

Python 3 | idle

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

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Launcher Mail\_footfall\_prediction\_skeleton.py

```
svs_regressor
default model fitting...
Now tuning SVR(c=1.0, cache_size=200, coef0=0.0, degree=3, epsilon=0.1, gamma='scale',
kernel='rbf', max_iter=1, shrinking=True, tol=0.001, verbose=False).
Cross CV validation...
adaBoost
default model fitting...
Now tuning AdaBoostRegressor(base_estimator=None, learning_rate=1.0, loss='linear',
n_estimators=200, random_state=None).
```

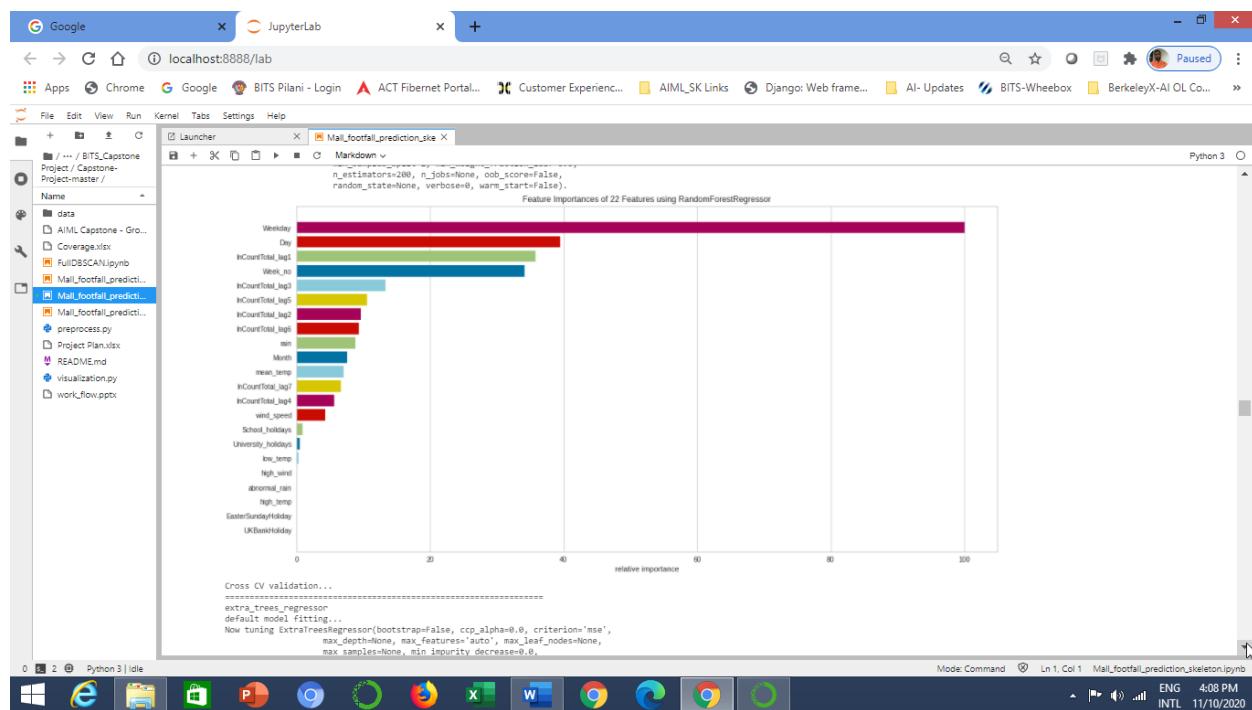
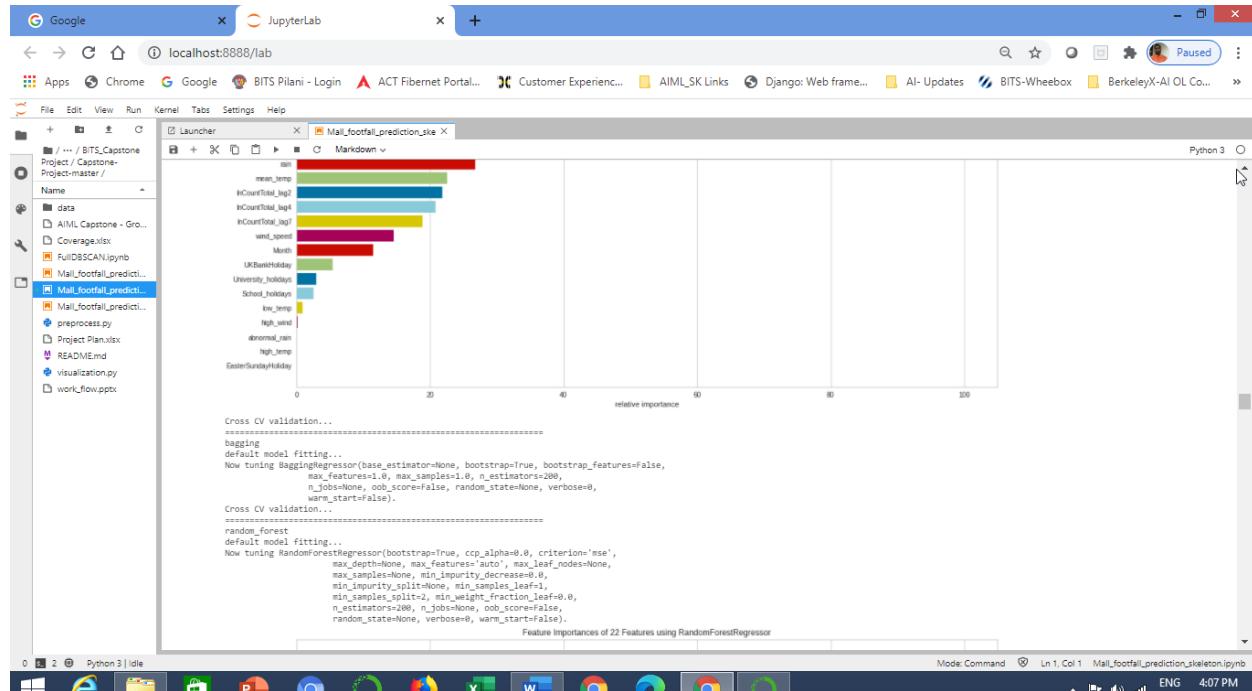
Feature Importances of 22 Features using AdaBoostRegressor

Cross CV validation...

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Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

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Launcher Mail\_footfall\_prediction\_skeleton.ipynb

```
extra_trees_regressor
default model fitting...
Now tuning ExtraTreesRegressor(ccp_alpha=0.0, criterion='mse',
                                max_depth=None, max_features='auto', max_leaf_nodes=None,
                                min_impurity_decrease=0.0,
                                min_samples=None, min_impurity_split=None, min_samples_leaf=1,
                                min_samples_split=2, min_weight_fraction_leaf=0.0,
                                n_estimators=200, n_jobs=None, oob_score=False,
                                random_state=None, verbose=0, warm_start=False).
```

Feature Importances of 22 Features using ExtraTreesRegressor

Feature	Relative Importance
Weekday	~75
Week_no	~45
Day	~35
HCountTotal_lag1	~25
HCountTotal_lag2	~20
Month	~18
HCountTotal_lag3	~15
HCountTotal_lag4	~15
wind_speed	~12
rain	~10
mean_temp	~10
HCountTotal_lag5	~8
HCountTotal_lag6	~8
University_holiday	~7
low_temp	~5
School_holiday	~5
high_wind	~3
high_rain	~3
abnormal_rain	~3
UKBankHoliday	~2
EasterSundayHoliday	~2

0 2 Python 3 | idle Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb ENGLISH 4:09 PM INTL 11/10/2020

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Launcher Mail\_footfall\_prediction\_skeleton.ipynb

```
Cross CV validation...
=====
gradient_boosting_regressor
default model fitting...
Now tuning GradientBoostingRegressor(alpha=0.9, ccp_alpha=0.0, criterion='friedman_mse',
                                         int=None, learning_rate=0.1, loss='ls', max_depth=3,
                                         max_features=None, max_leaf_nodes=None,
                                         min_impurity_decrease=0.0, min_impurity_split=None,
                                         min_samples=None, min_weight_fraction_leaf=0.0,
                                         n_estimators=200, n_iter_no_change=None, presort='deprecated',
                                         random_state=None, subsample=1.0, tol=0.0001,
                                         validation_fraction=0.1, verbose=0, warm_start=False).
```

Feature Importances of 22 Features using GradientBoostingRegressor

Feature	Relative Importance
Weekday	~75
Day	~45
Week_no	~35
HCountTotal_lag1	~25
HCountTotal_lag2	~20
Month	~18
HCountTotal_lag3	~15
HCountTotal_lag4	~15
wind_speed	~12
rain	~10
mean_temp	~10
HCountTotal_lag5	~8
HCountTotal_lag6	~8
University_holiday	~7
low_temp	~5
School_holiday	~5
high_wind	~3
high_rain	~3
abnormal_rain	~3
UKBankHoliday	~2
EasterSundayHoliday	~2

0 2 Python 3 | idle Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb ENGLISH 4:10 PM INTL 11/10/2020

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Launcher Mail\_footfall\_prediction\_ske

relative importance

Cross CV validation...

```
XGBoost
default model fitting...
Now tuning XGBRegressor(base_score=0.5, booster='gbtree', colsample_bytree=1,
colsample_bynode=1, colsample_bylevel=1, gamma=0,
importance_type='gain', learning_rate=0.1, max_delta_step=0,
max_depth=3, min_child_weight=1, missing=None, n_estimators=100,
n_jobs=1, random_state=42, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
seed=None, silent=None, subsample=1, verbosity=1)
```

Feature Importances of 22 Features using XGBRegressor

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Launcher Mail\_footfall\_prediction\_ske

relative importance

Nhg\_Wind  
abnormal\_rain  
UKBankHoliday

Cross CV validation...

```
LighGBMBooster
default model fitting...
Now tuning LGBMRegressor(boosting_type='gbdt', class_weight=None, colsample_bytree=1.0,
importance_type='split', learning_rate=0.1, max_depth=-1,
min_child_samples=20, min_child_weight=0, min_split_gain=0.0,
n_estimators=100, num_leaves=31, objective='regression',
random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
subsample=1.0, subsample_for_bin=200000, subsample_freq=0).
```

Cross CV validation...

```
CategoricalGradientBoosting
default model fitting...
Now tuning CatBoostRegressor object at 0x7f9271b12728.
Cross CV validation...
```

NeuralNet\_MLP
default model fitting...
Now tuning MLPRegressor(activation='relu', alpha=0.0001, batch\_size='auto', beta\_1=0.9,
beta\_2=0.99, early\_stopping=False, epsilon=1e-09,
hidden\_layer\_sizes=(100,), learning\_rate='constant',
learning\_rate\_init=0.001, max\_fun=150000, max\_iter=200,
momentum=0.9, n\_iter\_no\_change=10, nesterov\_momentum=True,
power\_t=0.5, random\_state=None, shuffle=True, solver='adam',
tol=1e-05, validation\_fraction=0.1, verbose=False,
warm\_start=False),
Cross CV validation...

Python 3

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 Python 3 | idle ENG 4:11 PM INTL 11/10/2020

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Launcher Mail\_footfall\_prediction\_ske

```
[37]: loss_of_outlier_removed.sort_values('mape').head(25)
```

	Model_type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
39	random_forest_Outliers_Removed_shuffled_auto_model	11.69754	0.004138	-0.013269	0.042451	-0.009544	0.064315	0.481009	0.445163	-1913.193796	-1824.059573
36	bagging_Outliers_Removed_shuffled_auto_model	11.759516	0.004164	-0.013649	0.044665	-0.010641	0.064529	0.477538	0.441452	-1910.814239	-1821.826516
45	gradient_boosting_regressor_Outliers_Removed...	12.215195	0.004682	-0.017509	0.047274	-0.021555	0.068422	0.412601	0.372030	-1779.803372	
52	LightGradientBoosting_Outliers_Removed_shuffl...	12.250588	0.004421	-0.016596	0.046592	-0.017115	0.066490	0.445303	0.409991	-1889.440826	-1800.252903
40	random_forest_Outliers_Removed_shuffled_rsmo...	12.332604	0.004447	-0.013993	0.045452	-0.008571	0.066687	0.442208	0.403488	-1887.326676	-1798.138753
54	CategoricalGradientBoosting_Outliers_Removed...	12.398471	0.004391	-0.013883	0.046761	-0.010403	0.066265	0.449098	0.411038	-1891.856969	-1802.497767
55	GradientBoosting_CrossValidation_Outliers_Rem...	12.382344	0.004390	-0.015885	0.047120	-0.016408	0.066259	0.449149	0.411103	-1891.924883	-1802.736960
43	extra_trees_regressor_Outliers_Removed_shuffl...	12.415550	0.004390	-0.015885	0.047120	-0.016408	0.066259	0.449149	0.411103	-1891.924883	-1802.736960
42	extra_trees_regressor_Outliers_Removed_shuffl...	12.459072	0.004411	-0.015203	0.047198	-0.014595	0.065620	0.448477	0.408248	-1800.197171	-1801.092048
48	XGBoost_Outliers_Removed_shuffled_default_mod...	12.577377	0.004815	-0.017852	0.048143	-0.018971	0.069387	0.395919	0.354195	-1859.996953	-1769.805708
46	gradient_boosting_regressor_Outliers_Removed...	12.632322	0.004775	-0.016552	0.048288	-0.014528	0.069103	0.400854	0.359472	-1861.922079	-1772.734156
44	extra_trees_regressor_Outliers_Removed_shuffl...	12.641993	0.004270	0.001854	0.044996	0.358338	0.065291	0.425421	NaN	NaN	NaN
51	LightGradientBoosting_Outliers_Removed_shuffl...	12.709911	0.004599	-0.014983	0.048127	-0.014091	0.067815	0.422976	0.388121	-1873.352644	-1766.164721
37	bagging_Outliers_Removed_shuffled_rsmo_model	13.049147	0.004865	-0.016208	0.049521	-0.008872	0.069732	0.389888	0.347748	-1858.4446934	-1766.259011
56	CategoricalGradientBoosting_Outliers_Removed...	13.132737	0.004550	-0.000036	0.047296	0.327100	0.067242	0.385288	NaN	NaN	NaN
41	random_forest_Outliers_Removed_shuffled_cros...	13.233996	0.004534	0.001688	0.048930	0.391672	0.067086	0.398068	NaN	NaN	NaN
34	adaBoost_Outliers_Removed_shuffled_rsmo_model	13.259735	0.004769	-0.010863	0.046260	0.003871	0.069060	0.401605	0.360274	-1862.39705	-1773.181782
49	XGBoost_Outliers_Removed_shuffled_rsmo_model	13.259941	0.004772	-0.010815	0.049262	0.006114	0.069082	0.401215	0.359887	-1862.137055	-1772.949132
47	gradient_boosting_regressor_Outliers_Removed...	13.445203	0.004540	0.001265	0.047983	0.353404	0.067219	0.393039	NaN	NaN	NaN
35	adaBoost_Outliers_Removed_shuffled_crosrsmo_m...	13.917584	0.004873	0.002081	0.047995	0.382242	0.068452	0.351508	NaN	NaN	NaN
12	huberRegressor_Outliers_Removed_shuffled_rsmo...	13.974179	0.005750	-0.018115	0.053593	-0.016486	0.075828	0.278562	0.228733	-1795.612475	-1706.424552
13	huberRegressor_Outliers_Removed_shuffled_rsmo...	14.025577	0.005841	-0.016998	0.053099	-0.021893	0.076425	0.267163	0.216548	-1700.015758	-1700.627835
38	bagging_Outliers_Removed_shuffled_rsmo_model	14.217424	0.004912	0.000908	0.050483	0.4198927	0.069815	0.358690	NaN	NaN	NaN
10	elastic_net_Outliers_Removed_shuffled_rsmo_mod...	14.331859	0.005460	-0.018112	0.053930	-0.014345	0.075165	0.291117	0.242155	-1801.880079	-1712.692158

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Launcher Mail\_footfall\_prediction\_ske

```
[38]: reg_models = regression_model(data.outliers_removed.dataset_type='OutliersRemoved_extra_features_shuffled')
for loss_normal_df in reg_models.train_and_evaluate_models(ifCross_validation=False,shuffle=True, extra_input_features=True):
    loss_values_all_models.append(loss_normal_df)
```

Defined 28 models

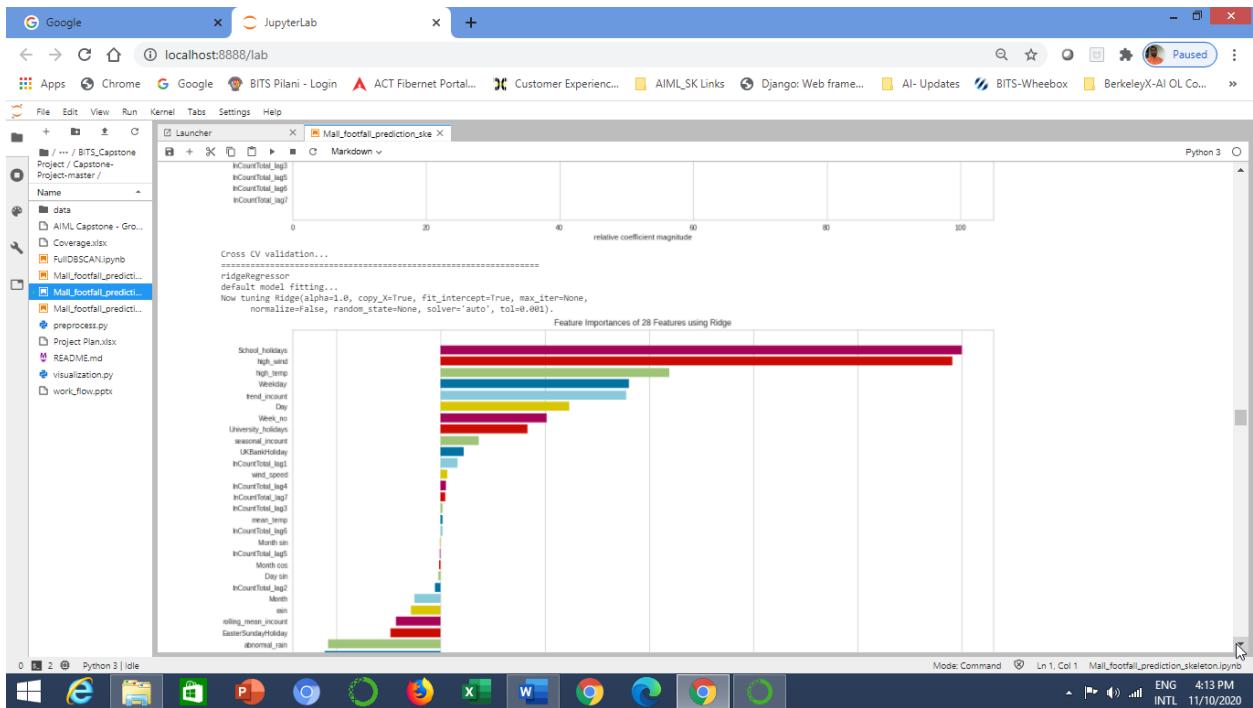
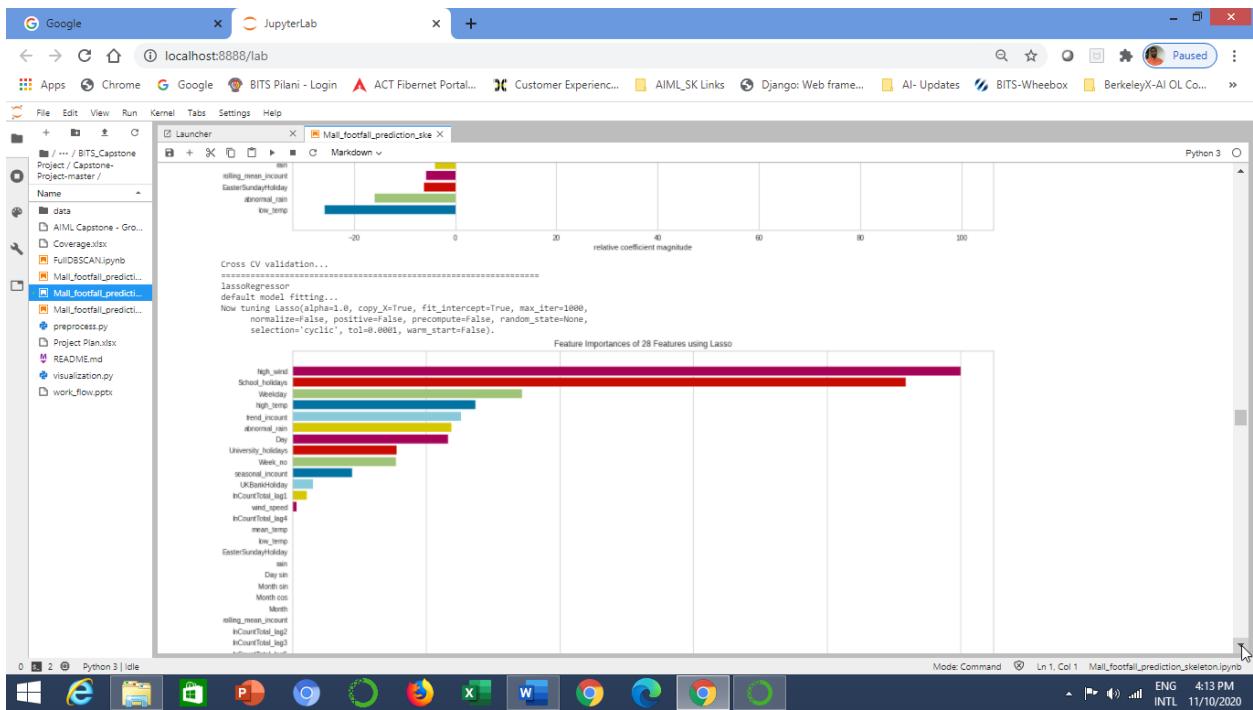
LinearRegression

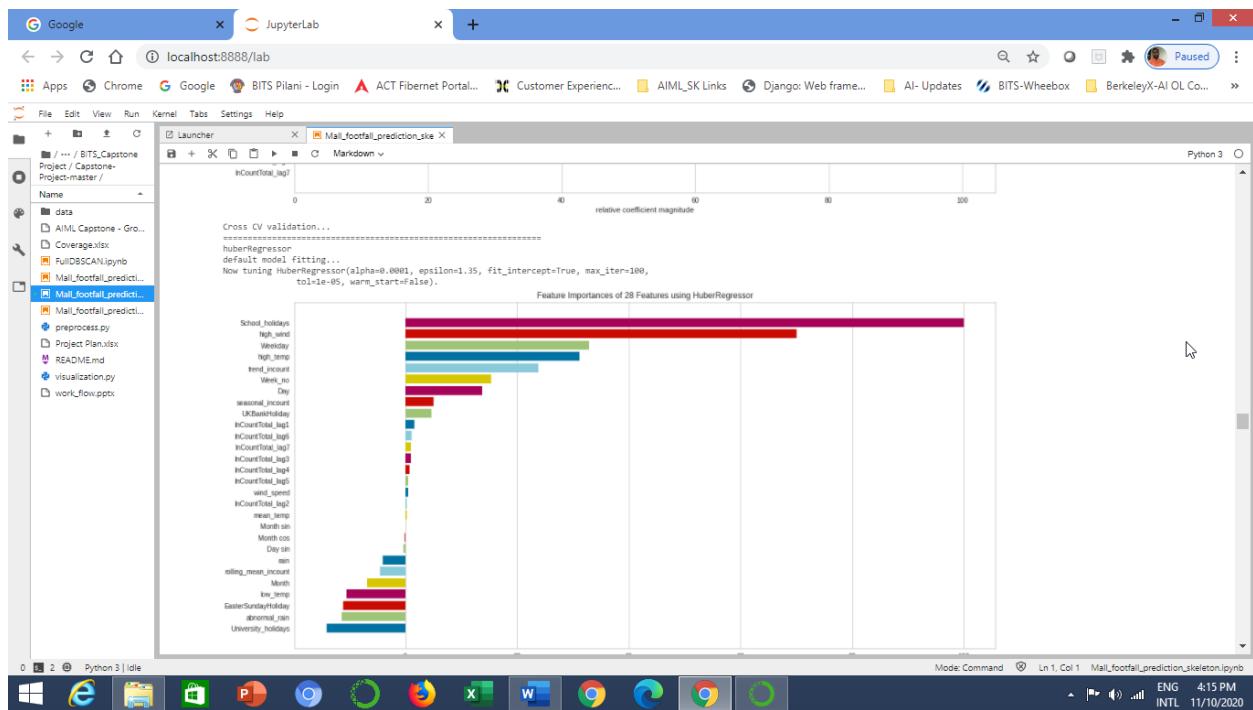
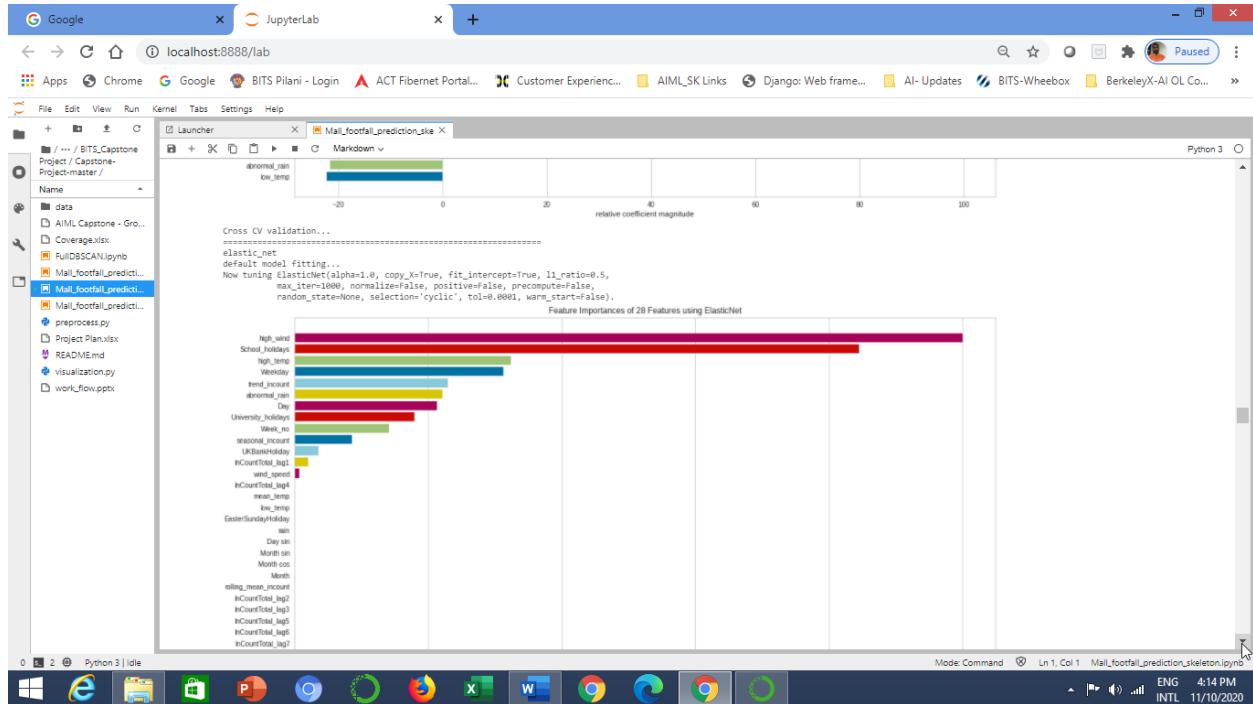
default model fitting...

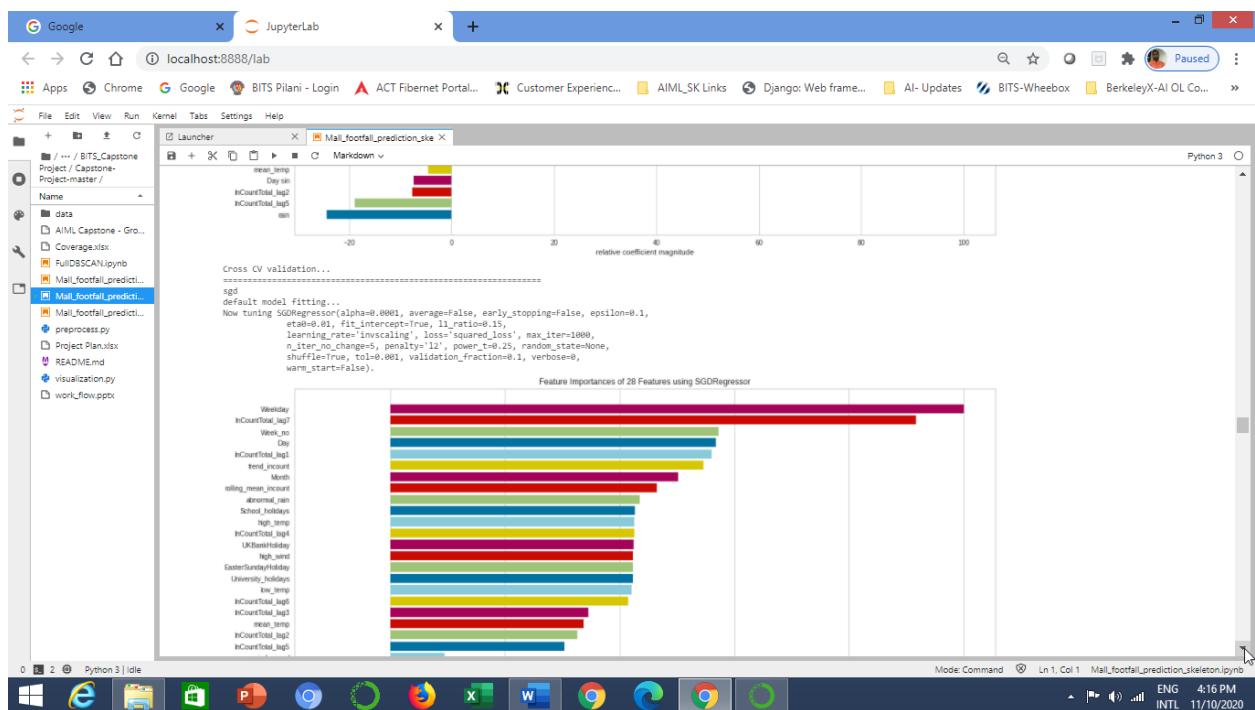
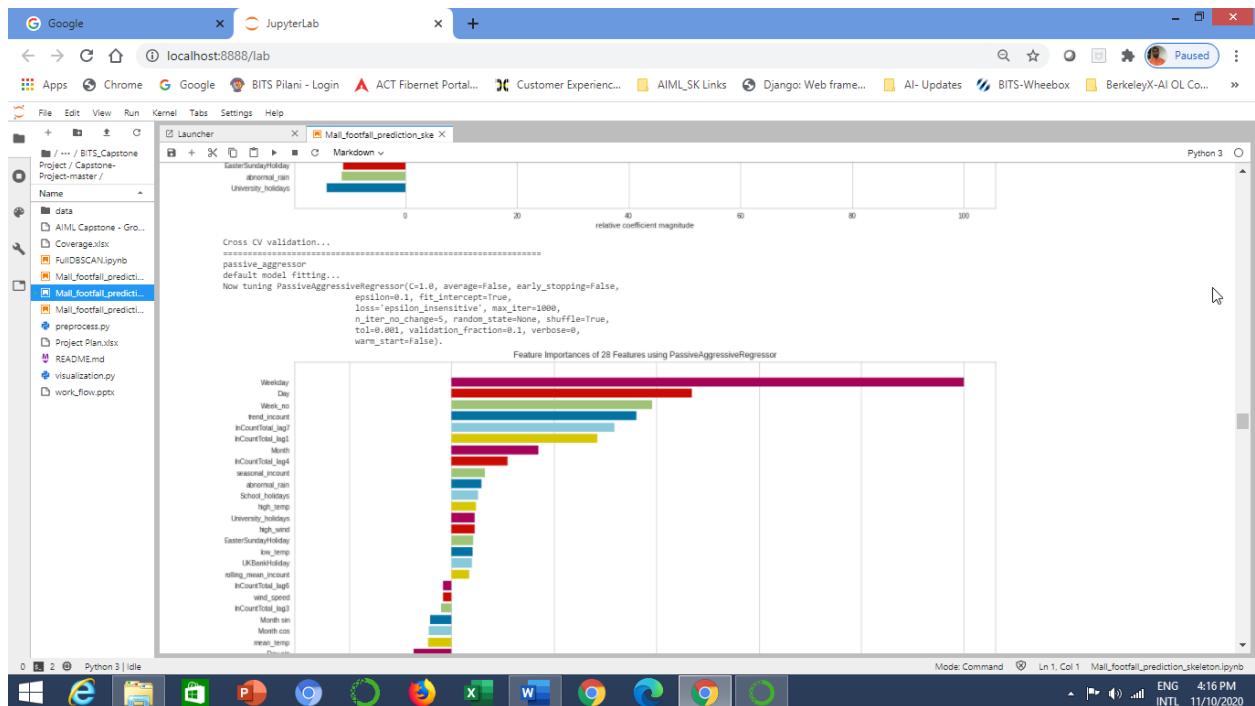
Now tuning LinearRegression(copy\_X=True, fit\_intercept=True, n\_jobs=None, normalize=False).

Feature Importances of 28 Features using LinearRegression

Feature	Importance (approx.)
Nth\_wend	0.15
School\_holiday	0.12
Nth\_temp	0.10
Weekday	0.08
IsEnd\_incount	0.07
Day	0.06
University\_holiday	0.05
Week\_no\_seasonal\_incount	0.04
UKBankholiday	0.03
ICCountTotal\_lag1	0.02
wind\_speed	0.01
ICCountTotal\_lag1\_lag1	0.005
ICCountTotal\_lag1\_lag2	0.005
mean\_temp	0.005
ICCountTotal\_lag2	0.005
Month\_sin	0.005
Month\_cos	0.005
Day\_sin	0.005
ICCountTotal\_lag3	0.005
Month\_min	0.005
min	0.005
end\_mean\_incount	0.005







The screenshot shows a JupyterLab interface with a Python 3 kernel. The notebook displays two bar charts showing feature importance and a log of cross-validation and model fitting steps.

**Feature Importance Plot:**

relative coefficient magnitude

Feature	Relative Coefficient Magnitude
incomesinerage	~85
mean_temp	~35
InCountTotal_lag1	~30
InCountTotal_lag2	~25
wind_spread	~15
Month cos	~10
Month sin	~5
seasonal_incount	~2
Day sin	~1
sin	~0.5

**Cross CV validation...**

default model fitting...

Now tuning KNeighborsRegressor(algorithm='auto', leaf\_size=30, metric='minkowski', metric\_params=None, n\_jobs=None, n\_neighbors=7, p=2, weights='uniform').

**Cross CV validation...**

decisiontree\_regressor

default model fitting...

Now tuning DecisionTreeRegressor(ccp\_alpha=0.0, criterion='mse', max\_depth=None, max\_features=None, max\_leaf\_nodes=None, min\_impurity\_decrease=0.0, min\_impurity\_split=None, min\_samples\_leaf=1, min\_samples\_split=2, min\_weight\_fraction\_leaf=0, presort='deprecated', random\_state=None, splitter='best').

Feature Importances of 28 Features using DecisionTreeRegressor

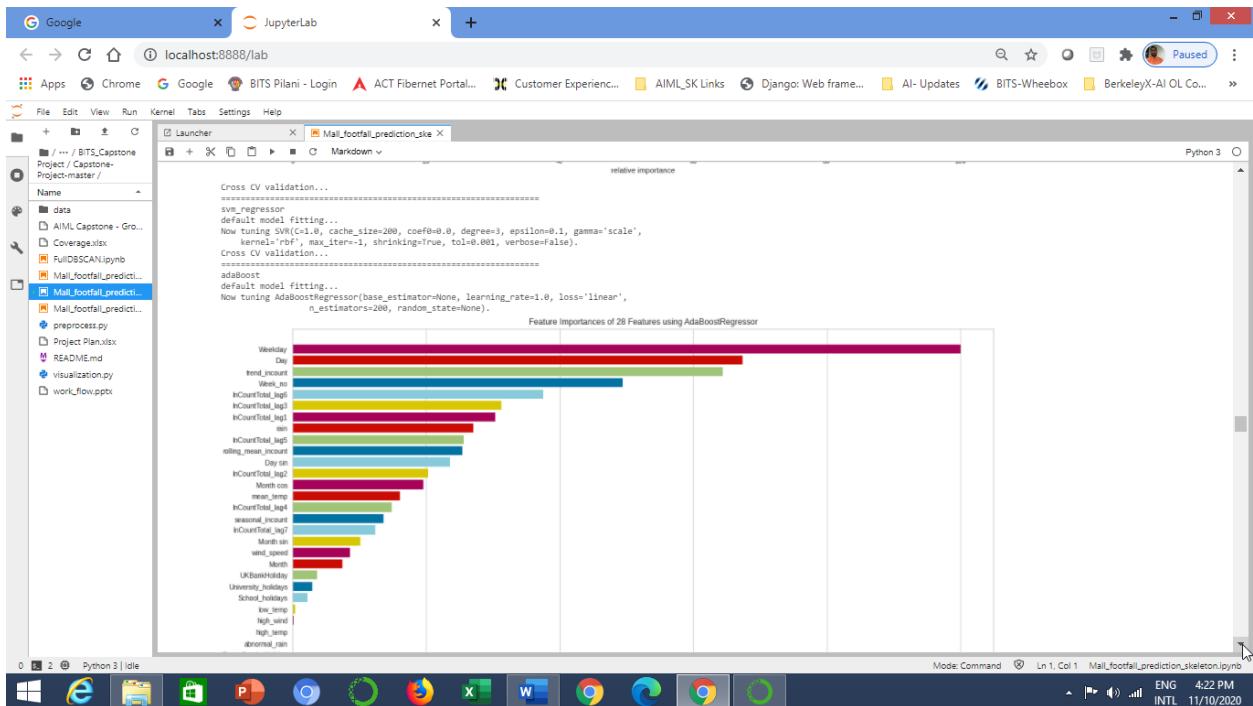
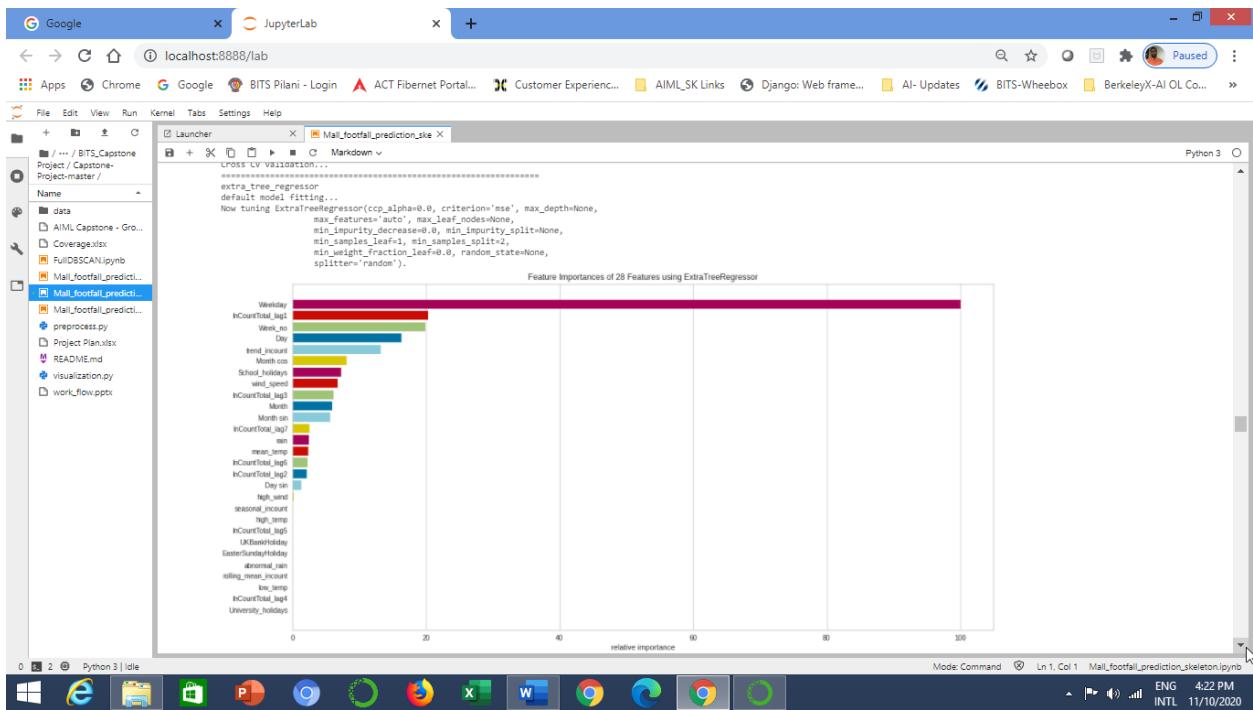
Feature	Importance
Weekday	~95
Day	~70
InCountTotal_lag1	~50
trend_incount	~25
decentrality	~20
InCountTotal_lag2	~15
InCountTotal_lag3	~10
Month sin	~5
InCountTotal_lag4	~2
InCountTotal_lag5	~1

The figure shows a JupyterLab interface with a Python 3 kernel. The code in the editor is as follows:

```
weights='uniform').  
Cross CV validation.....  
*****  
default model fitting..  
Now tuning DecisionTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,  
max_features=None, max_leaf_nodes=None,  
min_impurity_decrease=0.0, min_impurity_split=None,  
min_samples_leaf=1, min_samples_split=2,  
presort='deprecated',  
random_state=None, splitter='best').
```

Below the code is a horizontal bar chart titled "Feature Importances of 28 Features using DecisionTreeRegressor". The x-axis represents the importance score, ranging from 0 to 100. The y-axis lists the features. The most important feature is "Weekday", with an importance of approximately 95. Other significant features include "Day", "InCountTotal\_1sat", and "InCountTotal\_1sun".

Feature	Importance
Weekday	~95
Day	~85
InCountTotal_1sat	~75
Week_no	~45
InCountTotal_1apr	~35
InCountTotal_1jul	~30
sun	~25
Mon	~15
InCountTotal_1ap4	~10
InCountTotal_1ap3	~10
InCountTotal_1ap2	~10
InCountTotal_1ap1	~10
Month	~10
rolling_mean_incount	~10
Month cos	~10
mean_temp	~10
Day sin	~10
seasonal_index	~10
abnormal_rain	~10
high_temp	~10
low_temp	~10
EasterSundayHoliday	~10
UniversityHolidays	~10
School_holidays	~10
UKBankerholidays	~10
wind_speed	~10



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Launcher Mail\_footfall\_prediction\_skeleton.py

```

abnormal_rain
EasterSundayHoliday
relative importance 0 20 40 60 80 100

Cross CV validation...
=====
bagging
default model fitting.
Now tuning BaggingRegressor(base_estimator=None, bootstrap=True, bootstrap_features=False,
max_features=1.0, max_samples=1.0, n_estimators=200,
n_jobs=None, oob_score=False, random_state=None, verbose=0,
warm_start=False).

Cross CV validation...
=====
random_forest
default model fitting...
Now tuning RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
max_depth=None, max_features='auto', max_leaf_nodes=None,
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_leaf=0.0,
n_estimators=200, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False).

Feature Importances of 28 Features using RandomForestRegressor
Weekday Day
InCountTotal_lag1
Wind_ae
Wind_in
InCountTotal_lag3
InCountTotal_lag2
InCountTotal_lag4
InCountTotal_lag5
seasonal_incount
Month_cos
rolling_mean_incount
rolling_mean_lag1
Day_sin
InCountTotal_lag6

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3

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```

Day_sin
InCountTotal_lag1
wind_speed
Month
North_east
School_holiday
University_holiday
high_temp
high_wend
low_temp
Unemployment
abnormal_rain
EasterSundayHoliday
relative importance 0 20 40 60 80 100

Cross CV validation...
=====
extra_trees_regressor
default model fitting...
Now tuning ExtraTreeRegressor(bootstrap=False, ccp_alpha=0.0, criterion='mse',
max_depth=None, max_features='auto', max_leaf_nodes=None,
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_leaf=0.0,
n_estimators=200, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False).

Feature Importances of 28 Features using ExtraTreesRegressor
Weekday Day
InCountTotal_lag1
Wind_ae
Day
InCountTotal_lag3
InCountTotal_lag2
InCountTotal_lag4
InCountTotal_lag5
Month
Month_cos
seasonal_incount
InCountTotal_lag6
rolling_mean_incount

```

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Python 3

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Launcher Mail\_footfall\_prediction\_ske

Python 3

relative importance

Feature	Relative Importance
Weekday	~70
Week_no	~25
InCountTotal_1gpt	~15
rolling_mean_incount	~10
wind_speed	~10
InCountTotal_1m	~10
Month_sin	~10
mean_temp	~10
Day_sin	~10
School_holidays	~10
InCountTotal_1w	~10
InCountTotal_1gpt	~10
low_temp	~10
University_holidays	~10
high_wind	~5
high_temp	~5
abnormal_rain	~5
UKbankholiday	~5
EasternSundayHoliday	~5

Cross CV validation...

```
gradient_boosting_regressor
default model fitting...
Now tuning GradientBoostingRegressor(alpha=0.9, ccp_alpha=0.0, criterion='friedman_mse',
init=None, learning_rate=0.1, loss='ls', max_depth=3,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, n_estimators=200,
n_iter_no_change=None, prefit=False, random_state=42,
random_state=None, subsample=1.0, validate_presubsample_size=True,
validation_fraction=0.1, verbose=0, warm_start=False).
```

Feature Importances of 28 Features using GradientBoostingRegressor

Feature	Relative Importance
Weekday	~65
Week_no	~25
InCountTotal_1gpt	~15
Day	~10
Week_no	~10

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Launcher Mail\_footfall\_prediction\_ske

Python 3

relative importance

Feature	Relative Importance
Weekday	~65
Week_no	~25
InCountTotal_1gpt	~15
InCountTotal_1m	~10
InCountTotal_1w	~10
rolling_mean_incount	~10
mean_temp	~10
InCountTotal_1gpt	~10
Day_sin	~10
InCountTotal_1gpt	~10
Month_cos	~10
seasonal_incount	~10
wind_speed	~10
Month_sin	~10
School_holidays	~10
University_holidays	~10
low_temp	~10
abnormal_rain	~5
UKbankholiday	~5
high_temp	~5
high_wind	~5
EasternSundayHoliday	~5

Cross CV validation...

```
XGBRegressor
default model fitting...
Now tuning XGBRegressor(base_score=0.5, booster='gbtree', colsample_bylevel=1,
colsample_bynode=1, colsample_bytree=1, gamma=0,
learning_rate=0.1, max_delta_step=0,
max_depth=3, min_child_weight=1, missing=None, n_estimators=100,
n_jobs=1, nthread=None, objective='reg:squarederror',
random_state=42, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
seed=None, silent=None, subsample=1, verbosity=1).
```

Feature Importances of 28 Features using XGBRegressor

Feature	Relative Importance
Weekday	~65
Week_no	~25

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Launcher Mail\_footfall\_prediction\_ske

```
XGBoost
default model fitting...
Now tuning XGBRegressor(base_score=0.5, booster='gbtree', colsample_bytree=1,
colsample_bynode=1, colsample_bylevel=1,
importance_type='gain', learning_rate=0.1, max_delta_step=0,
max_depth=3, min_child_weight=1, missing=None, n_estimators=100,
n_jobs=-1, nthread=None, objective='regression',
random_state=42, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
seed=None, silent=None, subsample=1, verbosity=1).
```

Feature Importances of 28 Features using XGBRegressor

Feature	Relative Importance
Weekday	~95
InCountTotal_lag1	~45
InCountTotal_lag2	~42
InCountTotal_lag3	~40
InCountTotal_lag4	~38
InCountTotal_lag5	~35
InCountTotal_lag6	~32
InCountTotal_lag7	~30
InCountTotal_lag8	~28
InCountTotal_lag9	~25
Month	~25
Month_sin	~22
Month_cos	~20
Day_sin	~18
Day_cos	~15
Year	~15
Year_sin	~12
Year_cos	~10
HighTemp	~10
LowTemp	~8
WindSpeed	~8
University_holiday	~5
EasterSundayholiday	~3
UKbankholiday	~3
Autumn_winter	~3
HighTemp	~2
LowTemp	~2
WindSpeed	~2

Python 3

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```
Cross CV validation...
=====
LightGBM default model fitting...
Now tuning LGBMRegressor(boosting_type='gbdt', class_weight=None, colsample_bytree=1.0,
importance_type='split', learning_rate=0.1, max_depth=-1,
min_child_samples=20, min_child_weight=0.001, min_split_gain=0.0,
n_estimators=100, num_leaves=31, objective='regression',
random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
subsample=1.0, subsample_for_bin=200000, subsample_freq=0).
Cross CV validation...
=====
Categorical gradient boosting
default model fitting...
Now tuning catboost.core.CatBoostRegressor object at 0x7f9271800cf8>.
Cross CV validation...
=====
NeuralNet_MLP
default model fitting...
Now tuning MLPRegressor(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
beta_2=0.99, early_stopping=False, epsilon=1e-05,
hidden_layer_sizes=(100,), learning_rate='constant',
learning_rate_init=0.001, max_fun=150000, max_iter=200,
momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True,
power_t=0.5, random_state=None, shuffle=True, solver='adam',
tol=0.0001, validation_fraction=0.1, verbose=False,
warm_start=False).
Cross CV validation...
```

Python 3

Solve PC issues: 1 message

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```
[39]: loss_df = pd.concat([loss_each_model for loss_each_model in loss_values_all_models], axis=0).reset_index(drop=True)

[40]: loss_df.sort_values('mape').head(25)
```

	Model_type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
299	random_forest_Outliers_Removed_shuffled_default	11.697544	0.004136	-0.013269	0.044251	-0.009544	0.064315	0.481009	0.445163	-1913.19796	-1824.005873
290	bagging_Outliers_Removed_shuffled_default	11.759516	0.004164	-0.013640	0.044665	-0.010641	0.064529	0.477538	0.441452	-1910.814239	-1821.626516
230	bagging_Outliers_Removed_default	11.815070	0.004205	-0.013330	0.044784	-0.009610	0.064849	0.472342	0.435897	-1907.281280	-1818.093357
233	random_forest_Outliers_Removed_default	11.878325	0.004131	-0.012771	0.044793	-0.007844	0.064274	0.481667	0.445666	-1913.646660	-1824.457873
248	CategoricalGradientBoosting_Outliers_Removed	11.991476	0.004348	-0.012474	0.045388	-0.007304	0.065937	0.454493	0.416015	-1895.404784	-1806.216861
234	random_forest_Outliers_Removed	12.020228	0.004409	-0.015048	0.045648	-0.012178	0.066401	0.446783	0.408373	-1890.394484	-1801.206541
358	extra_trees_regressor_Outliers_Removed_extra	12.128559	0.003951	-0.011822	0.043080	3.487159	0.062699	0.472110	NaN	NaN	NaN
239	gradient_boosting_regressor_Outliers_Removed	12.205760	0.004710	-0.017822	0.047270	-0.021842	0.068652	0.408985	0.368164	-1866.80007	-1777.612084
350	bagging_Outliers_Removed_extra_features	12.209038	0.004184	-0.005161	0.044888	0.015016	0.064681	0.475084	0.428332	-1997.141257	-1784.686920
299	gradient_boosting_regressor_Outliers_Removed	12.213195	0.004682	-0.017509	0.047274	-0.021555	0.068422	0.4121601	0.372030	-1868.991395	-1779.803372
306	LightGradientBoosting_Outliers_Removed_shuffle	12.250636	0.004421	-0.016398	0.045992	-0.017115	0.066490	0.445303	0.406991	-1889.440826	-1800.252903
246	LightGradientBoosting_Outliers_Removed	12.250636	0.004421	-0.016398	0.045992	-0.017115	0.066490	0.445303	0.406991	-1889.440826	-1800.252903
353	random_forest_Outliers_Removed_extra_features	12.262066	0.004228	-0.004918	0.045078	0.015555	0.065021	0.449551	0.422509	-1903.398286	-1780.949348
294	random_forest_Outliers_Removed_shuffled	12.332504	0.004447	-0.013993	0.046429	-0.008571	0.066887	0.442008	0.403468	-1887.326678	-1798.138753
308	CategoricalGradientBoosting_Outliers_Removed	12.338471	0.004391	-0.013688	0.046781	-0.010404	0.066263	0.449059	0.411038	-1891.885690	-1802.697767
309	CategoricalGradientBoosting_Outliers_Removed	12.382344	0.004682	-0.016381	0.047490	-0.016135	0.067916	0.421260	0.381287	-1874.292851	-1785.104928
297	extra_trees_regressor_Outliers_Removed_shuffle	12.415150	0.004390	-0.015855	0.047120	-0.016406	0.065359	0.449149	0.411103	-1891.924883	-1802.736960
357	extra_trees_regressor_Outliers_Removed	12.400017	0.004233	-0.009966	0.045960	0.004382	0.065750	0.457590	0.409487	-1888.437816	-1772.983479
296	extra_trees_regressor_Outliers_Removed	12.455072	0.004412	-0.015203	0.047198	-0.014598	0.066420	0.446477	0.408245	-1890.197171	-1801.009248
356	extra_trees_regressor_Outliers_Removed_extra	12.461586	0.004330	-0.009301	0.046153	0.003958	0.065801	0.456744	0.408566	-1884.881484	-1772.427127
236	extra_trees_regressor_Outliers_Removed_default	12.529236	0.004402	-0.015246	0.047466	-0.014811	0.066348	0.447709	0.409563	-1890.992657	-1801.804734
237	extra_trees_regressor_Outliers_Removed	12.538173	0.004136	-0.015710	0.047578	-0.015818	0.066494	0.445370	0.405138	-1888.372207	-1790.139791

Mode Command L1 Col 1 Mail\_footfall\_prediction\_skeleton.py

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```
[40]:
```

	Model_type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
299	random_forest_Outliers_Removed_shuffled	11.697544	0.004136	-0.013269	0.044251	-0.009544	0.064315	0.481009	0.445163	-1913.19796	-1824.005873
290	bagging_Outliers_Removed_shuffled	11.759516	0.004164	-0.013640	0.044665	-0.010641	0.064529	0.477538	0.441452	-1910.814239	-1821.626516
230	bagging_Outliers_Removed	11.815070	0.004205	-0.013330	0.044784	-0.009610	0.064849	0.472342	0.435897	-1907.281280	-1818.093357
233	random_forest_Outliers_Removed	11.878325	0.004131	-0.012771	0.044793	-0.007844	0.064274	0.481667	0.445666	-1913.646660	-1824.457873
248	CategoricalGradientBoosting_Outliers_Removed	11.991476	0.004348	-0.012474	0.045388	-0.007304	0.065937	0.454493	0.416015	-1895.404784	-1806.216861
234	random_forest_Outliers_Removed	12.020228	0.004409	-0.015048	0.045648	-0.012178	0.066401	0.446783	0.408373	-1890.394484	-1801.206541
358	extra_trees_regressor_Outliers_Removed	12.128559	0.003951	-0.011822	0.043080	3.487159	0.062699	0.472110	NaN	NaN	NaN
239	gradient_boosting_regressor_Outliers_Removed	12.205760	0.004710	-0.017822	0.047270	-0.021842	0.068652	0.408985	0.368164	-1866.80007	-1777.612084
350	bagging_Outliers_Removed_extra_features	12.209038	0.004184	-0.005161	0.044888	0.015016	0.064681	0.475084	0.428332	-1997.141257	-1784.686920
299	gradient_boosting_regressor_Outliers_Removed	12.213195	0.004682	-0.017509	0.047274	-0.021555	0.068422	0.4121601	0.372030	-1868.991395	-1779.803372
306	LightGradientBoosting_Outliers_Removed	12.250636	0.004421	-0.016398	0.045992	-0.017115	0.066490	0.445303	0.406991	-1889.440826	-1800.252903
246	LightGradientBoosting_Outliers_Removed	12.250636	0.004421	-0.016398	0.045992	-0.017115	0.066490	0.445303	0.406991	-1889.440826	-1800.252903
353	random_forest_Outliers_Removed_extra_features	12.262066	0.004228	-0.004918	0.045078	0.015555	0.065021	0.449551	0.422509	-1903.398286	-1780.949348
294	random_forest_Outliers_Removed_shuffled	12.332504	0.004447	-0.013993	0.046429	-0.008571	0.066887	0.442008	0.403468	-1887.326678	-1798.138753
308	CategoricalGradientBoosting_Outliers_Removed	12.338471	0.004391	-0.013688	0.046781	-0.010404	0.066263	0.449059	0.411038	-1891.885690	-1802.697767
309	CategoricalGradientBoosting_Outliers_Removed	12.382344	0.004682	-0.016381	0.047490	-0.016135	0.067916	0.421260	0.381287	-1874.292851	-1785.104928
297	extra_trees_regressor_Outliers_Removed	12.415150	0.004390	-0.015855	0.047120	-0.016406	0.065359	0.449149	0.411103	-1891.924883	-1802.736960
357	extra_trees_regressor_Outliers_Removed	12.400017	0.004233	-0.009966	0.045960	0.004382	0.065750	0.457590	0.409487	-1888.437816	-1772.983479
296	extra_trees_regressor_Outliers_Removed	12.455072	0.004412	-0.015203	0.047198	-0.014598	0.066420	0.446477	0.408245	-1890.197171	-1801.009248
356	extra_trees_regressor_Outliers_Removed	12.461586	0.004330	-0.009301	0.046153	0.003958	0.065801	0.456744	0.408566	-1884.881484	-1772.427127
236	extra_trees_regressor_Outliers_Removed_default	12.529236	0.004402	-0.015246	0.047466	-0.014811	0.066346	0.447709	0.409563	-1890.992657	-1801.804734
237	extra_trees_regressor_Outliers_Removed	12.538173	0.004435	-0.015710	0.047578	-0.015818	0.066594	0.445370	0.405138	-1888.327207	-1799.139284
354	random_forest_Outliers_Removed_extra_features	12.533699	0.004409	-0.008588	0.045977	0.014153	0.066397	0.446851	0.397795	-1878.438706	-1785.984368
302	XGBoost_Outliers_Removed	12.577377	0.004815	-0.017662	0.048142	-0.018978	0.069387	0.395919	0.354196	-1858.993651	-1789.805708
242	XGBoost_Outliers_Removed	12.577377	0.004815	-0.017862	0.048142	-0.018978	0.069387	0.395919	0.354196	-1858.993651	-1789.805708

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7.1.5 Model training with PCA features

```
[41]: reg_models = regression_model(princdf,dataset_type='PCA_features')
... loss_pca_df = reg_models.train_and_evaluate_models(#FECross_validation=False,pcafeatures=True)
loss_values_all_models.append(loss_pca_df)
```

Defined 28 models

linear\_regression

default model fitting..

Now tuning LinearRegression(copy\_X=True, fit\_intercept=True, n\_jobs=None, normalize=False).

Feature Importances of 12 Features using LinearRegression

Feature	Importance (approx.)
0	-95
1	-75
2	-65
3	-55
4	-45
5	-35
6	-25
7	-15
8	85
9	75
10	65
11	55

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

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Cross CV validation...

lassoRegressors

default model fitting..

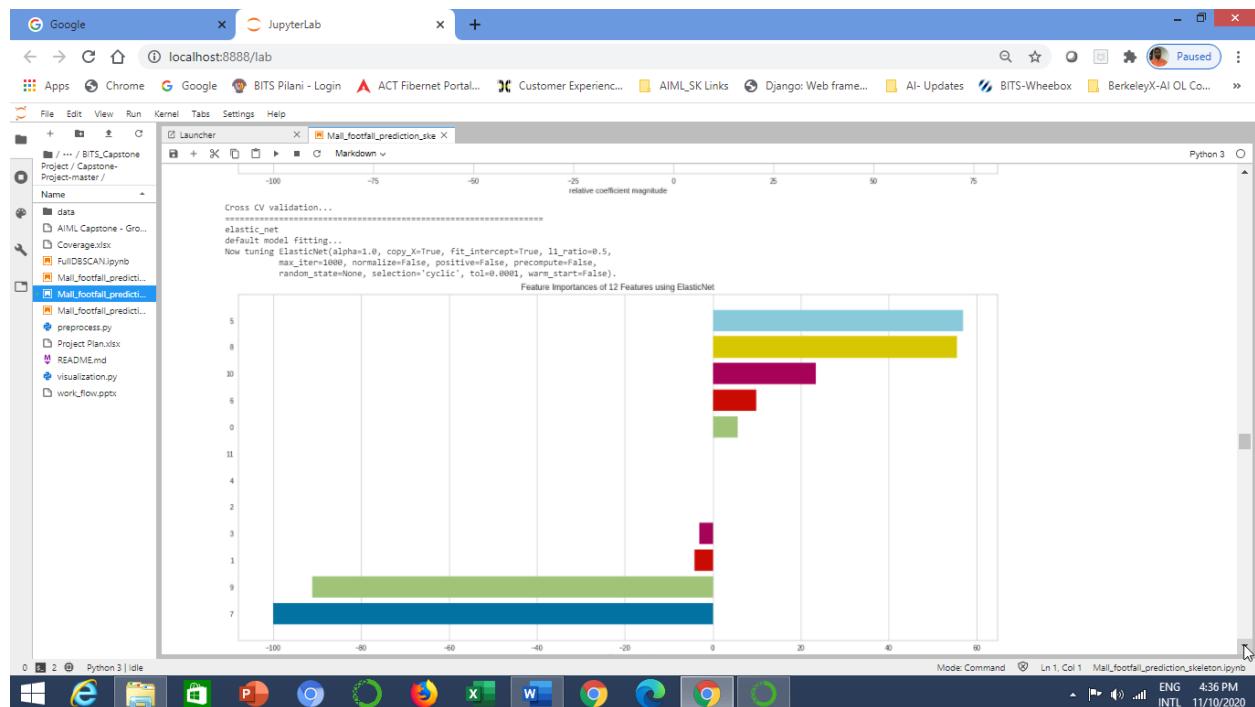
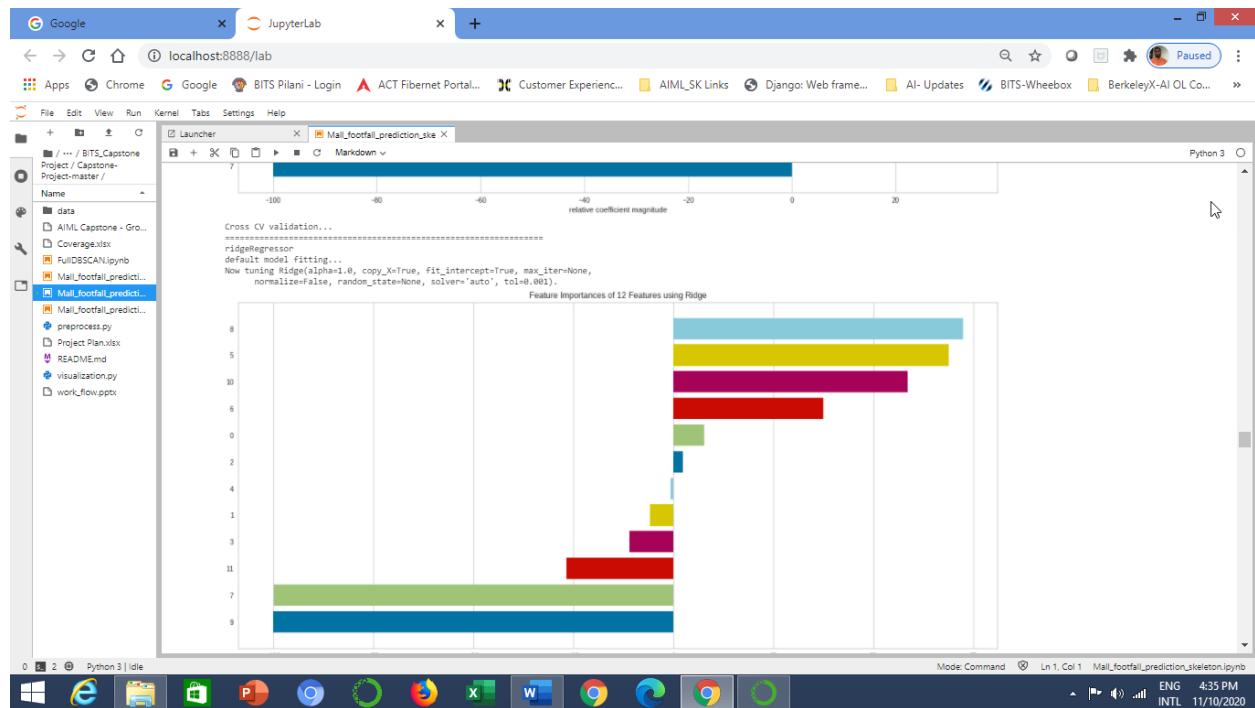
Now tuning Lasso(alpha=0.0, copy\_X=True, fit\_intercept=True, max\_iter=1000, normalize=False, positive=False, precompute=False, random\_state=None, selection='cyclic', tol=0.0001, warm\_start=False).

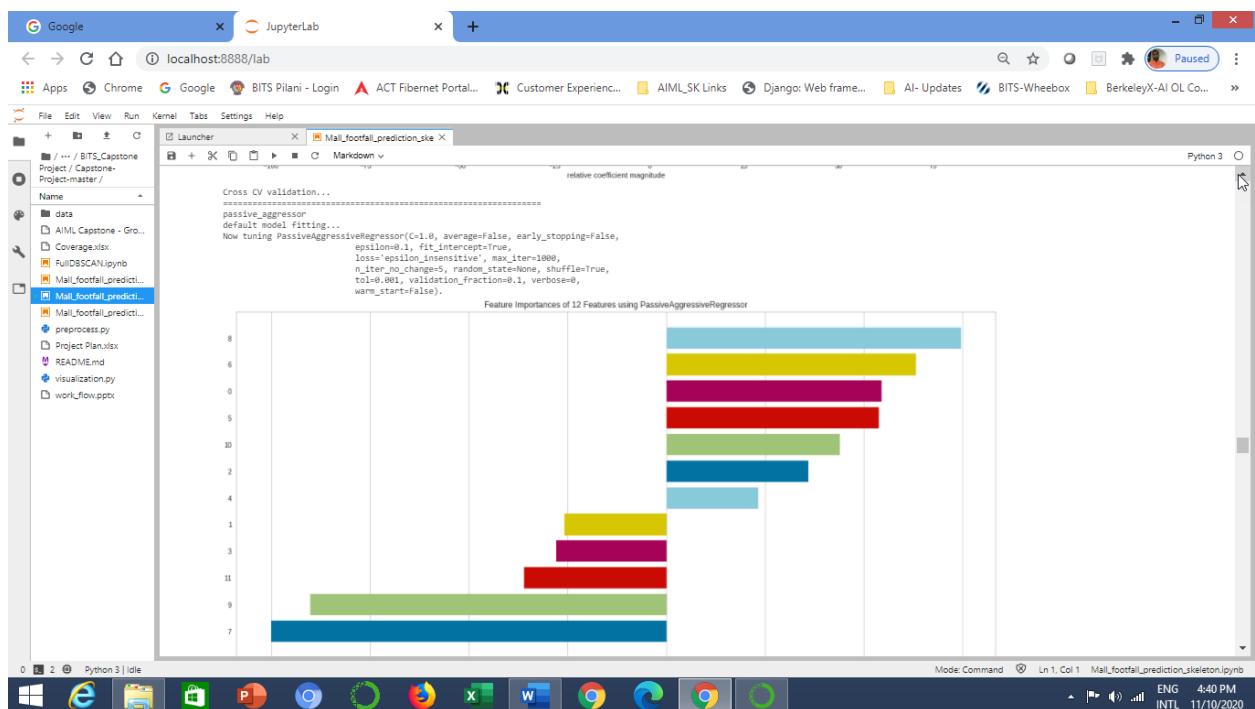
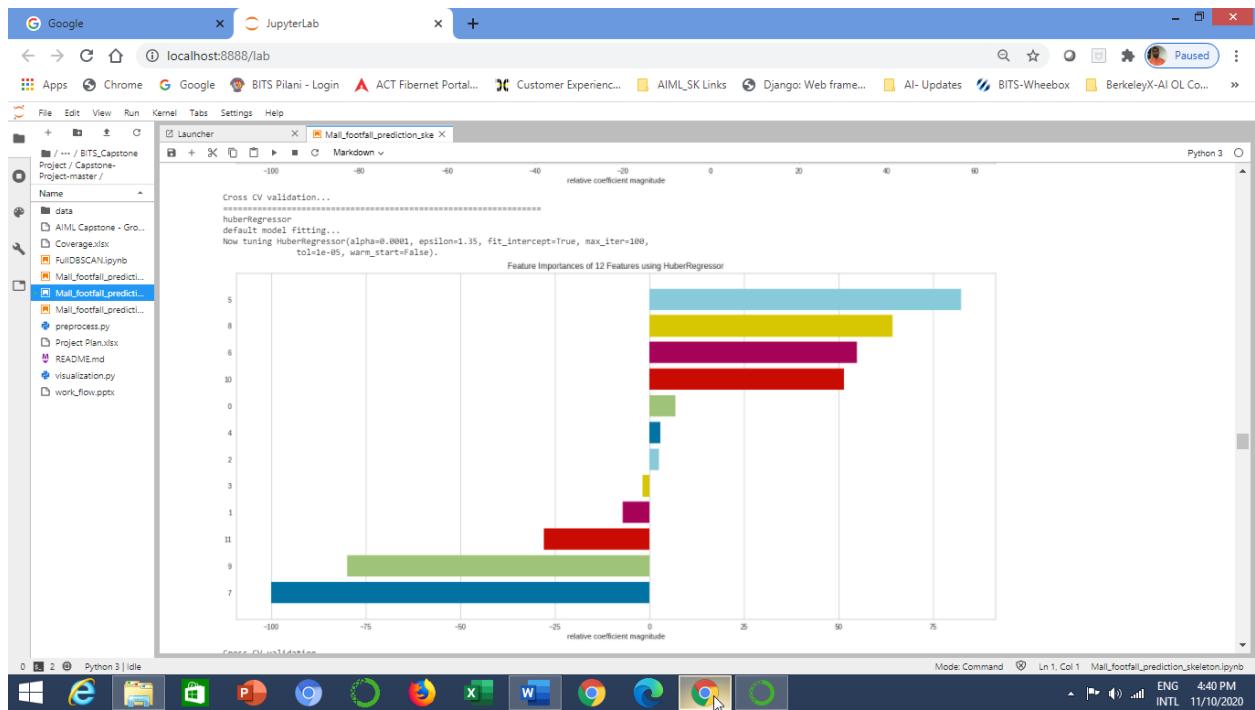
Feature Importances of 12 Features using Lasso

Feature	Importance (approx.)
0	-95
1	-75
2	-65
3	-55
4	-45
5	-35
6	-25
7	-15
8	85
9	75
10	65
11	55

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

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The screenshot shows a JupyterLab interface with a code cell containing SGDRegressor fitting logs and a bar chart titled "Feature Importances of 12 Features using SGDRegressor".

```
Cross CV validation...
481
default model fitting...
Now tuning SGDRegressor(alpha=0.0001, average=False, early_stopping=False, epsilon=0.1,
eta0=0.01, fit_intercept=True, l1_ratio=0.15,
loss='squared_loss', max_iter=1000,
n_iter_no_change=5, penalty='l2', power_t=25, random_state=None,
shuffle=True, tol=0.001, validation_fraction=0.1, verbose=0,
warm_start=False).
```

Feature Importances of 12 Features using SGDRegressor

Feature	Importance
1	-0.05
2	-0.08
3	-0.05
4	-0.05
5	-0.05
6	-0.05
7	-0.05
8	0.05
9	0.05
10	0.05
11	0.05
12	0.05

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Launcher Mail\_footfall\_prediction\_skeleton.ipynb Python 3

```
Cross CV validation...
=====
knn_regressor
default model fitting...
Now tuning KNeighborsRegressor(algorithm='auto', leaf_size=30, metric='minkowski',
                               metric_params=None, n_jobs=None, n_neighbors=7, p=2,
                               weights='uniform').
Cross CV validation...
=====
decisiontree_regressor
default model fitting...
Now tuning DecisionTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,
                                   max_features=None, max_leaf_nodes=None,
                                   min_impurity_decrease=0.0, min_impurity_split=None,
                                   min_samples_leaf=1, min_samples_split=2,
                                   min_weight_fraction_leaf=0.0, presort='deprecated',
                                   random_state=None, splitter='best')
    Feature Importances of 12 Features using DecisionTreeRegressor
    7
    8
    6
    5
    4
    9
    3
    11
    10
    2
```

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Launcher Code Python 3

relative importance

Cross CV validation...

```
=====
extra_tree_regression
default model fitting...
Now tuning ExtraTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,
max_features='auto', max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=2,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, random_state=None,
splitter='random').
```

Feature Importances of 12 Features using ExtraTreeRegressor

Feature	Importance
6	~95
7	~45
11	~25
5	~15
4	~10
10	~10

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Launcher Code Python 3

relative importance

Cross CV validation...

```
=====
svr_regression
default model fitting...
Now tuning SVR(C=1.0, cache_size=200, coef0=0.0, degree=3, epsilon=0.1, gamma='scale',
kernel='rbf', max_iter=1, shrinking=True, tol=0.001, verbose=False).
Cross CV validation...
```

adaBoost

default model fitting...

```
Now tuning AdaBoostRegressor(base_estimator=None, learning_rate=1.0, loss='linear',
n_estimators=200, random_state=None).
```

Feature Importances of 12 Features using AdaBoostRegressor

Feature	Importance
9	~95
8	~55
7	~45
5	~40

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```

relative importance
7
6
5
4
3
2
1
0
10
11
9
8
7
6
5
4
3
2
1
0
relative importance
Cross CV validation...
=====
bagging
default model fitting...
Now tuning BaggingRegressor(base_estimator=None, bootstrap=True, bootstrap_features=False,
    n_estimators=10, max_samples=1.0, n_estimators=200,
    oob_score=True, error_score=0, random_state=None, verbose=0,
    warm_start=False)
Cross CV validation...
=====
random_forest
default model fitting...
Now tuning RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
    max_depth=None, max_features='auto', max_leaf_nodes=None,
    max_samples=None, min_impurity_decrease=0.0,
    min_impurity_split=None, min_samples_leaf=1,
    min_samples_split=2, min_weight_fraction_leaf=0.0,
    n_estimators=200, n_jobs=None, oob_score=True,
    random_state=None, verbose=0, warm_start=False)

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3

Feature	Relative Importance
7	~50
6	~45
5	~40
4	~35
3	~30
2	~25
1	~20
10	~38
11	~35
9	~32
8	~30
7	~28
6	~25
5	~22
4	~20
3	~18
2	~15
1	~12

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Launcher Mail\_footfall\_prediction\_ske

```

random_forest
default model fitting...
Now tuning RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
    max_depth=None, max_features='auto', max_leaf_nodes=None,
    max_samples=None, min_impurity_decrease=0.0,
    min_impurity_split=None, min_samples_leaf=1,
    min_samples_split=2, min_weight_fraction_leaf=0.0,
    n_estimators=200, n_jobs=None, oob_score=True,
    random_state=None, verbose=0, warm_start=False)

```

Feature Importances of 12 Features using RandomForestRegressor

Feature	Importance
9	~95
7	~80
6	~75
5	~65
4	~55
3	~40
2	~35
1	~25
10	~38
11	~32
8	~30
7	~28
6	~25
5	~22

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

Python 3

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Project / Capstone Project-master / Name

data Coverage.xlsx FullDBSCAN.ipynb Mail\_footfall\_prediction.ipynb Mail\_footfall\_prediction.py preprocess.py Project Plan.xlsx README.md visualization.py workflow.pptx

Cross CV validation...  
extra\_trees\_regressor  
default model fitting.  
Now tuning extraTreesRegressor(bootstrap=False, ccp\_alpha=0.0, criterion='mse', max\_depth=None, max\_features='auto', max\_leaf\_nodes=None, max\_samples=None, min\_impurity\_decrease=0.0, min\_imbalance\_split=None, min\_samples\_leaf=1, min\_samples\_split2=None, min\_weight\_fraction\_leaf=0.0, n\_estimators=200, n\_jobs=-1, oob\_score=False, random\_state=None, verbose=0, warm\_start=False).  
Feature Importances of 12 Features using ExtraTreesRegressor

Feature Index	Importance Score (approx.)
1	0.95
2	0.50
3	0.60
4	0.45
5	0.80
6	0.90
7	0.85
8	0.75
9	0.65
10	0.70
11	0.55
0	0.40

Python 3 | idle Mode: Command Ln 1, Col 1 - Mail\_footfall\_prediction\_skeleton.ipynb

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+ Launcher Mail\_footfall\_prediction\_ske

Cross CV Validation...  
gradient\_boosting\_regressor  
default model fitting...  
Now tuning GradientBoostingRegressor(...)  
int=None, learning\_rate=0.1, loss='ls', max\_depth=3,  
max\_features=None, max\_leaf\_nodes=None,  
min\_impurity\_decrease=0.0, min\_impurity\_split=None,  
min\_samples\_leaf=1, min\_samples\_split=2,  
min\_weight\_fraction\_leaf=0.0, n\_estimators=200,  
n\_iter\_no\_change=None, presort='deprecated',  
random\_state=None, subsample=1.0, tol=0.0001,  
validation\_fraction=0.1, verbose=0, warm\_start=False).

relative importance

Feature Importances of 12 Features using GradientBoostingRegressor

Feature Index	Importance Value
0	~9.2
1	~7.8
2	~6.5
3	~4.8
4	~3.5
5	~3.2
6	~2.8
7	~2.5
8	~2.2
9	~1.8
10	~1.5
11	~1.2

Python 3

Mode: Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 1 2 3 4 5 6 7 8 9 10 11 12

Windows Start Task View Home Back Forward Stop Refresh

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The screenshot shows a Jupyter Notebook interface with a sidebar containing project files and a main workspace displaying code and a bar chart.

**Code Cell Content:**

```
Now tuning GradientBoostingRegressor(alpha=0.9, ccp_alpha=0.0, criterion='friedman_mse',
                                         init=None, learning_rate=0.1, loss='ls', max_depth=3,
                                         max_features=None, max_leaf_nodes=None,
                                         min_impurity_decreased=0.01, min_impurity_split=None,
                                         min_leaf_node_size=1, min_weight_fraction_leaf=0.0,
                                         min_weight_fraction_leaf=0.0, n_estimators=200,
                                         n_iter_no_change=None, presort='deprecated',
                                         random_state=None, subsample=1.0, tol=0.001,
                                         validation_fraction=0.1, verbose=0, warm_start=False).
```

**Feature Importances Bar Chart:**

A horizontal bar chart titled "Feature Importances of 12 Features using GradientBoostingRegressor". The x-axis is labeled "relative importance" and ranges from 0 to 300. The y-axis lists indices from 0 to 11. The bars show the following approximate relative importances:

Index	Relative Importance (approx.)
0	42
1	18
2	24
3	40
4	26
5	90
6	105
7	150
8	55
9	280
10	45
11	38

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Cross CV validation...  
XGBoost  
default model fitted.  
Now tuning XGBRegressor(base\_score=0.5, booster='gbtree', colsample\_bytree=1,  
colsample\_bynode=1, colsample\_bytree=1, gamma=0,  
importance\_type='gain', learning\_rate=0.1, max\_delta\_step=0,  
max\_depth=3, min\_child\_weight=1, missing=None, n\_estimators=100,  
n\_jobs=-1, nthread=None, objective='reg:squarederror',  
random\_state=0, reg\_alpha=0, reg\_lambda=1, scale\_pos\_weight=1,  
seed=None, silent=None, subsample=1, verbosity=1).

Feature Importances of 12 Features using XGBRegressor

Feature Index	Relative Importance (approx.)
0	10
1	15
2	10
3	20
4	15
5	25
6	30
7	35
8	65
9	10
10	10
11	15

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```

Cross CV validation...
=====
LightGradientBoosting
default model fitting...
Now tuning LightGBMRegressor(booster_type='gbdt', class_weight=None, colsample_bytree=1.0,
                           importance_type='split', learning_rate=0.1, max_depth=-1,
                           min_child_samples=20, min_child_weight=0.001, min_split_gain=0.0,
                           n_estimators=100, n_jobs=-1, num_leaves=31, objective='none',
                           random_state=None, reg_alpha=0.0, reg_lambda=0.0, silent=True,
                           subsample=1.0, subsample_for_bin=200000, subsample_freq=0).
Cross CV validation...
=====
CategoricalGradientBoosting
default model fitting...
Now tuning CatBoost.core.CatBoostRegressor object at 0x7f92788567f0..
Cross CV validation...
=====
NeuralNet_MLP
default model fitting...
Now tuning MLPRegressor(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
                        beta_2=0.999, early_stopping=False, epsilon=1e-08,
                        hidden_layer_sizes=(100,), learning_rate='constant',
                        max_iter=200, max_n_ticks=100, max_wt_change=0.001,
                        momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True,
                        power_t=0.5, random_state=None, shuffle=True, solver='adam',
                        tol=0.0001, validation_fraction=0.1, verbose=False,
                        warm_start=False).
Cross CV validation...

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

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```
[42]: loss_pca_df.sort_values('mape').head(25)
```

	Model_type	mape	mse	me	mae	mpc	rmse	R2 score	Adj R2 score	AIC	BIC
52	LightGradientBoosting_PCA_features,rscv_model	26.804916	0.008612	-0.015611	0.005561	0.102041	0.092800	0.054554	0.019331	-1709.1030	-1658.71140
40	random_forest_PCA_features,rscv_model	26.835368	0.008381	-0.015167	0.004791	0.105417	0.091548	0.079734	0.046656	-1719.3399	-1668.441324
56	CategoricalGradientBoosting_PCA_features,prost	26.865394	0.007539	-0.001042	0.062709	14.151786	0.086643	0.140314	NaN	NaN	NaN
3	lasso_regressor_PCA_features,default,model	27.101444	0.009454	-0.016828	0.070133	0.091321	0.097234	-0.038100	-0.076548	-1675.363276	-1624.664611
9	elastic_net_PCA_features,default,model	27.101444	0.009454	-0.016828	0.070133	0.091321	0.097234	-0.038100	-0.076548	-1675.363276	-1624.664611
28	extra_tree_regressor_PCA_features,rscv_model	27.142544	0.009525	-0.015471	0.070481	0.087628	0.097598	-0.045948	-0.084584	-1672.649015	-1621.950349
4	lasso_regressor_PCA_features,rscv_model	27.222340	0.009127	-0.016749	0.068788	0.098939	0.095537	-0.0202183	-0.039300	-1688.215186	-1637.516952
41	random_forest_PCA_features,crossv,model	27.296745	0.007928	-0.000038	0.064920	14.412022	0.088687	0.107469	NaN	NaN	NaN
18	sgd_PCA_features,default,model	27.317651	0.009440	-0.019800	0.070197	0.091927	0.097160	-0.036519	-0.074908	-1675.919607	-1635.220941
19	sgd_PCA_features,rscv,model	27.390765	0.009357	-0.016777	0.069960	0.098622	0.096731	-0.027394	-0.064545	-1679.147158	-1628.484992
46	gradient_boosting_regressor_PCA_features,rs_cv	27.409875	0.008683	-0.019350	0.065653	0.109580	0.093182	0.046620	0.011300	-1706.439497	-1655.738281
53	LightGradientBoosting_PCA_features,crossv,model	27.573691	0.008416	-0.000672	0.067372	14.084049	0.091390	0.051430	NaN	NaN	NaN
16	passive_aggressor_PCA_features,rs_cv,model	27.598615	0.009198	-0.011805	0.069588	0.113127	0.095898	-0.009760	-0.047158	-1685.466418	-1634.767752
38	bagging_PCA_features,crossv,model	27.633249	0.008193	0.000283	0.066272	14.519286	0.090023	0.082698	NaN	NaN	NaN
37	bagging_PCA_features,rs_cv,model	27.637153	0.008648	-0.016884	0.066281	0.106902	0.092388	0.005056	0.015474	-1707.977925	-1657.279260
35	adaBoost_PCA_features,crossv,model	27.664349	0.008315	-0.002458	0.066746	13.774570	0.090916	0.059721	NaN	NaN	NaN
10	elastic_net_PCA_features,rs_cv,model	27.685808	0.008886	-0.013507	0.067969	0.111587	0.094265	0.024327	-0.011809	-1698.000584	-1647.301919
49	XGBboost_PCA_features,rs_cv,model	27.705307	0.008168	-0.006387	0.068264	0.134874	0.090367	0.030380	0.070151	-1728.833330	-1678.134654
17	passive_aggressor_PCA_features,crossv,model	27.764416	0.009088	-0.007489	0.069017	12.595700	0.094769	-0.016152	NaN	NaN	NaN
44	extra_tree_regressor_PCA_features,prost,model	27.848666	0.007777	0.001001	0.065921	15.200235	0.087905	0.120040	NaN	NaN	NaN
34	adaBoost_PCA_features,rs_cv,model	27.902777	0.009938	-0.012409	0.067614	0.115986	0.094333	0.018776	-0.017586	-1699.297333	-1645.231057
13	huberRegressors_PCA_features,rs_cv,model	28.074436	0.009017	-0.014827	0.068208	0.110247	0.094957	0.009945	-0.025723	-1692.659511	-1641.950546
12	huberRegressor_PCA_features,default,model	28.07604	0.009017	-0.014815	0.068209	0.110287	0.094958	0.009921	-0.026749	-1692.650476	-1641.951810

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Launcher Mail\_footfall\_prediction\_skeleton.py

```
[1]: huberRegressor_PCA_features_default_model 28.076404 0.009017 -0.014815 0.068209 0.110287 0.094958 0.009921 -0.026749 -1692.650476 -1641.951810
[2]: decisiontree_regressor_PCA_features_crosscv_model 28.136082 0.009324 -0.020682 0.068252 0.096787 0.096561 -0.023775 -0.061693 -1680.435018 -1629.736552
[3]: XGBBoost_PCA_features_crosscv_model 28.141559 0.007985 0.006062 0.066233 16.245185 0.089045 0.007405 NaN NaN NaN
```

7.1.6 Model training with Weekly data

```
[4]: reg_models = regression_model(data_weekly,dataset_type='weekly_data')
losses_weekly_df = reg_models.train_and_evaluate_models(NFEcross_validation=False,weekly_data=True)
```

Defined 20 models

linear\_regression  
default model fitting...

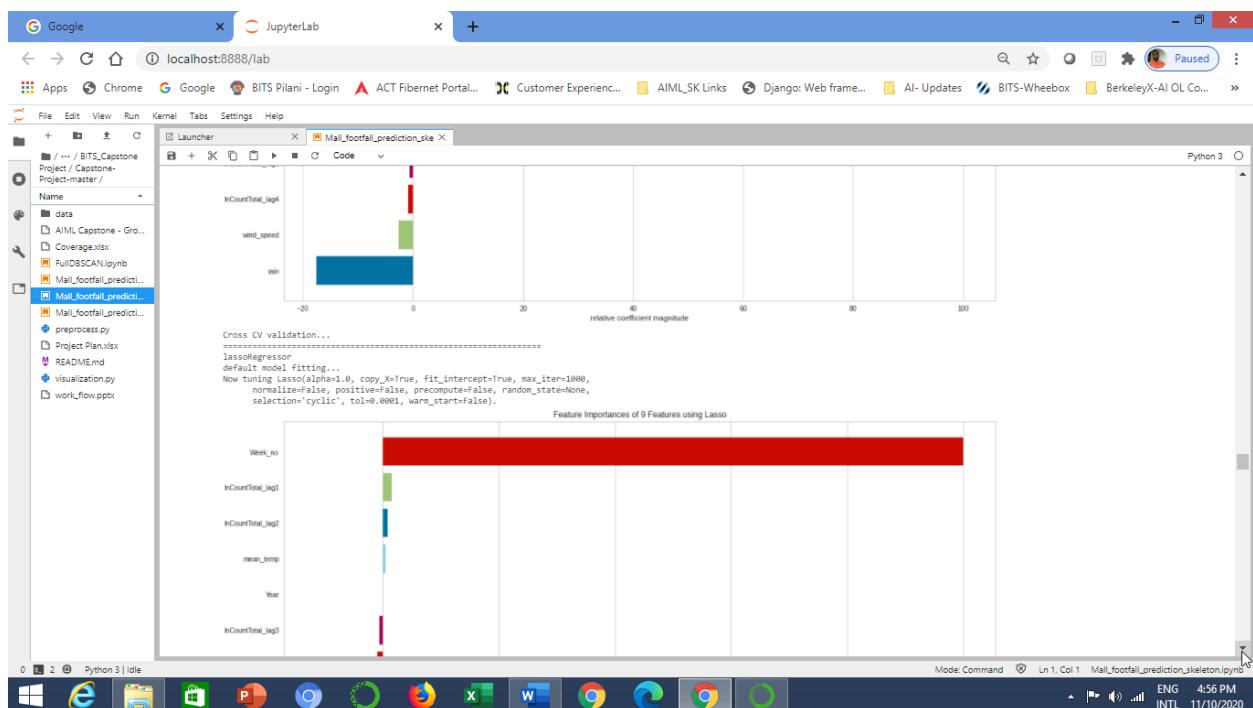
Now tuning linearregression(copy\_X=True, fit\_intercept=True, n\_jobs=None, normalize=False).

Feature Importances of 9 Features using LinearRegression

Feature	Importance
Week_No	~100
mean_temp	~5
InCountTotal_lag1	~5
InCountTotal_lag2	~5
Year	~5
InCountTotal_lag3	~5
InCountTotal_lag4	~5
nan	~5

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyb

0 2 Python 3 | idle ENG 4:55 PM INTL 11/10/2020



Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

relative coefficient magnitude

Cross CV validation...

```
ridgeRegress...  
default model fitting...  
Now tuning Ridge(alpha=1.0, copy_X=True, fit_intercept=True, max_iter=None,  
normalize=False, random_state=None, solver='auto', tol=0.001).
```

Feature Importances of 9 Features using Ridge

Feature	Relative Coefficient Magnitude
Week_no	~95
Year	~25
mean_temp	~5
InCountTotal_lag1	~5
InCountTotal_lag2	~5
InCountTotal_lag3	~5
InCountTotal_lag4	~5
wind_speed	~10
sun	~20

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 Python 3 | idle ENG 4:57 PM INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

relative coefficient magnitude

Cross CV validation...

```
elastic_net  
default model fitting...  
Now tuning ElasticNet(alpha=1.0, copy_X=True, fit_intercept=True, l1_ratio=0.5,  
max_iter=1000, normalize=False, positive=False, precompute=False,  
random_state=None, selection='cyclic', tol=0.0001, warm_start=False).
```

Feature Importances of 9 Features using ElasticNet

Feature	Relative Coefficient Magnitude
Week_no	~95
Year	~15
mean_temp	~5
InCountTotal_lag1	~5
InCountTotal_lag2	~5
inCountTotal_lag4	~5
wind_speed	~10
sun	~20

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

0 2 Python 3 | idle ENG 4:57 PM INTL 11/10/2020

The screenshot shows a Jupyter Notebook interface with several tabs open. The main notebook cell displays a bar chart titled "Feature Importances of 9 Features using HuberRegressor". The x-axis represents the relative coefficient magnitude from 0 to 100. The y-axis lists features: Week\_R3, Year, wnd\_speed, mean\_temp, lnCountTotal\_jag1, lnCountTotal\_jag2, and lnCountTotal\_jag3. The bars show that 'Week\_R3' has the highest importance (~95), followed by 'Year' (~30), and the others are much lower.

```
# Cross CV validation...
#####
# huberRegression
# default model fitting -
# Now tuning HuberRegressor(alpha=0.0001, epsilon=1.35, fit_intercept=True, max_iter=100,
# tol=1e-05, warm_start=False).
# Feature Importances of 9 Features using HuberRegressor
```

The screenshot shows a JupyterLab interface with a Python 3 kernel. On the left, a file tree displays various notebooks and scripts, including `Mail_footfall_prediction.ipynb`, which is currently selected. The main area contains a terminal window showing the execution of a script. The terminal output includes:

```
Cross CV validation...
=====
passive_aggressive
default model fitting...
Now tuning PassiveAggressiveRegressor(C=1.0, average=False, early_stopping=False,
epsilon=0.1, fit_intercept=True,
loss='epsilon_insensitive', max_iter=1000,
n_iter_no_change=5, random_state=None, shuffle=True,
tol=0.001, validation_fraction=0.1, verbose=0,
warm_start=False).
Feature Importances of 9 Features using PassiveAggressiveRegressor
```

Below the terminal, a horizontal bar chart titled "Feature Importances of 9 Features using PassiveAggressiveRegressor" is displayed. The x-axis represents the magnitude of the coefficient, ranging from -100 to 100. The y-axis lists the features: `inCountTotal_jag1`, `inCountTotal_jag2`, `mean_temp`, `Week_no`, `inCountTotal_jag3`, and `Year`. The bars show the following approximate values:

Feature	Importance (Relative Coefficient Magnitude)
<code>inCountTotal_jag1</code>	~95
<code>inCountTotal_jag2</code>	~80
<code>mean_temp</code>	~75
<code>Week_no</code>	~70
<code>inCountTotal_jag3</code>	~50
<code>Year</code>	~-40

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```

relative coefficient magnitude
Year
InCountTotal_lag4
wind_speed
min

```

Cross CV validation...

```

***** default model fitting...
Now tuning SGDRegressor(alpha=0.0001, average=False, early_stopping=False, epsilon=0.1,
eta0=0.01, fit_intercept=True, l1_ratio=0.15,
learning_rate='invscaling', loss='squared_loss', max_iter=1000,
n_iter_no_change=5, penalty='l2', power_t=0.25, random_state=None,
shuffle=True, tol=0.001, validation_fraction=0.1, verbose=0,
warm_start=False).

```

Feature Importances of 9 Features using SGDRegressor

```

relative coefficient magnitude
Week_NI
Year
mean_temp
InCountTotal_lag3
InCountTotal_lag2

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```

relative coefficient magnitude
wind_speed
InCountTotal_lag3
InCountTotal_lag2
min

```

Cross CV validation...

```

***** knn_regressor
default model fitting...
Now tuning KNeighborsRegressor(algorithm='auto', leaf_size=30, metric='minkowski',
metric_params=None, n_jobs=None, n_neighbors=7, p=2,
weights='uniform').

```

Cross CV validation...

```

***** decisiontree_regressor
default model fitting...
Now tuning DecisionTreeRegressor(ccp_alpha=0.0, criterion='mse', max_depth=None,
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, presort='deprecated',
random_state=None, splitter='best').

```

Feature Importances of 9 Features using DecisionTreeRegressor

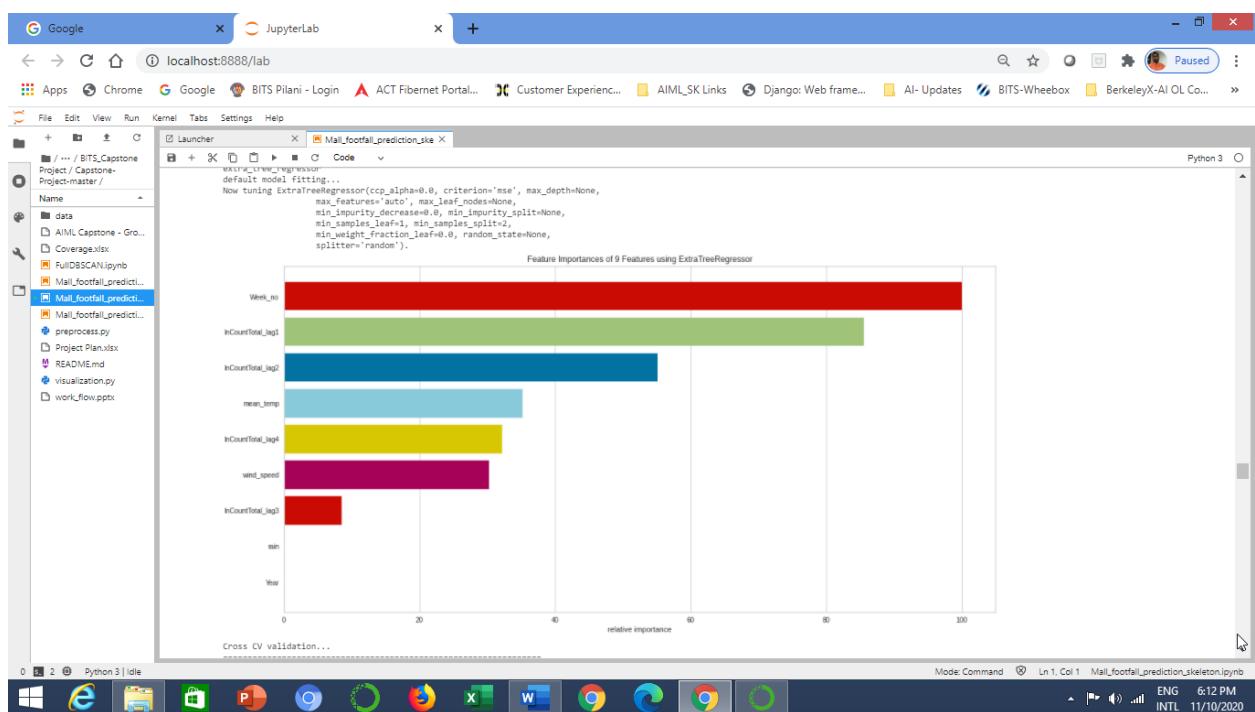
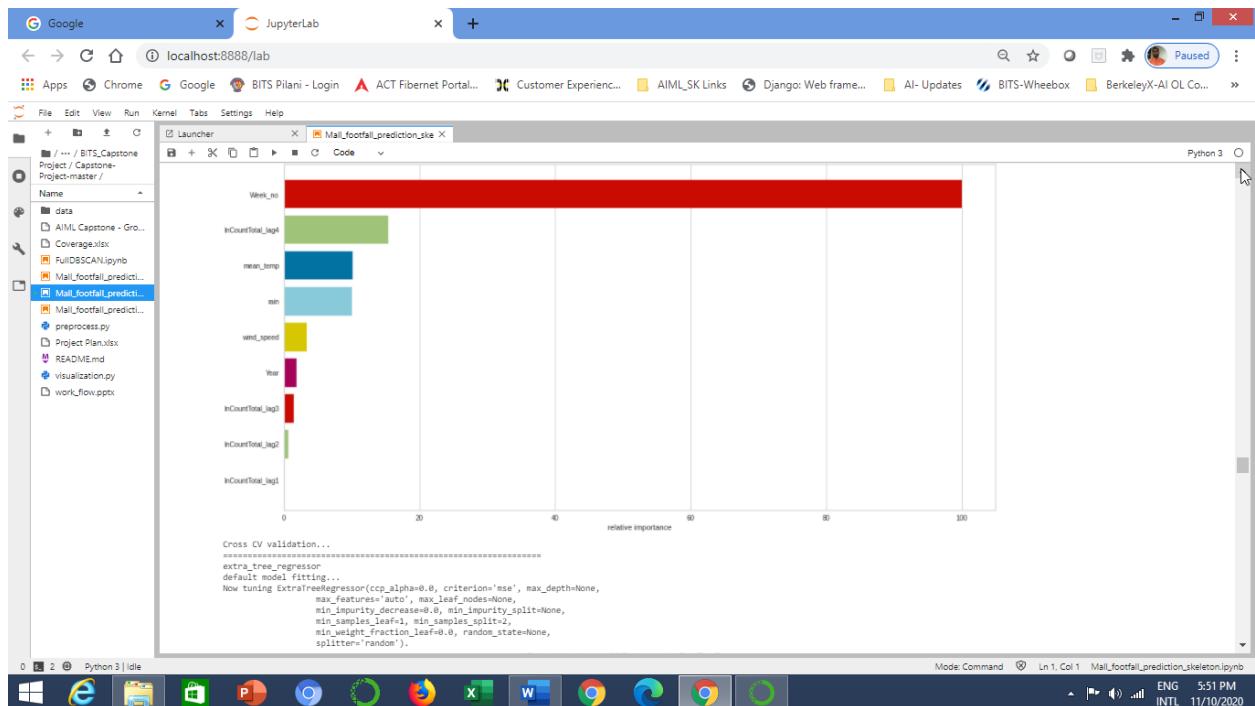
```

relative coefficient magnitude
Week_NI
InCountTotal_lag4

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3



The screenshot shows a JupyterLab interface with multiple tabs open. The main notebook cell displays the following code and output:

```
svs_regression
default model fitting...
Now tuning SVR(C=1.0, cache_size=200, coef0=0.0, degree=3, epsilon=0.1, gamma='scale',
kernel='rbf', max_iter=-1, shrinking=True, tol=0.001, verbose=False).
Cross CV validation...
adaBoost
default model fitting...
Now tuning AdaBoostRegressor(base_estimator=None, learning_rate=1.0, loss='linear',
n_estimators=200, random_state=None).
```

The output shows the feature importances for the AdaBoostRegressor model, represented by a horizontal bar chart titled "Feature Importances of 9 Features using AdaBoostRegressor".

Feature	Importance
wnd_spent	Very High (Red)
Week_ni	High (Green)
inCountTotal_jag4	Medium-High (Dark Blue)
inCountTotal_jag2	Medium-High (Light Blue)
inCountTotal_jag1	Medium-Low (Yellow)
sun	Low (Purple)
mean_timp	Low (Red)
inCountTotal_jag3	Low (Green)
Year	Very Low (Blue)

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Code Python 3

Project / Capstone Project-master /

Name

data

AI Capstone - Gro...

Coverage.xlsx

fullDBSCAN.ipynb

Mail\_footfall\_predict...

Mail\_footfall\_predict...

Mail\_footfall\_predict...

preprocess.py

Project Plan.xlsx

README.md

visualization.py

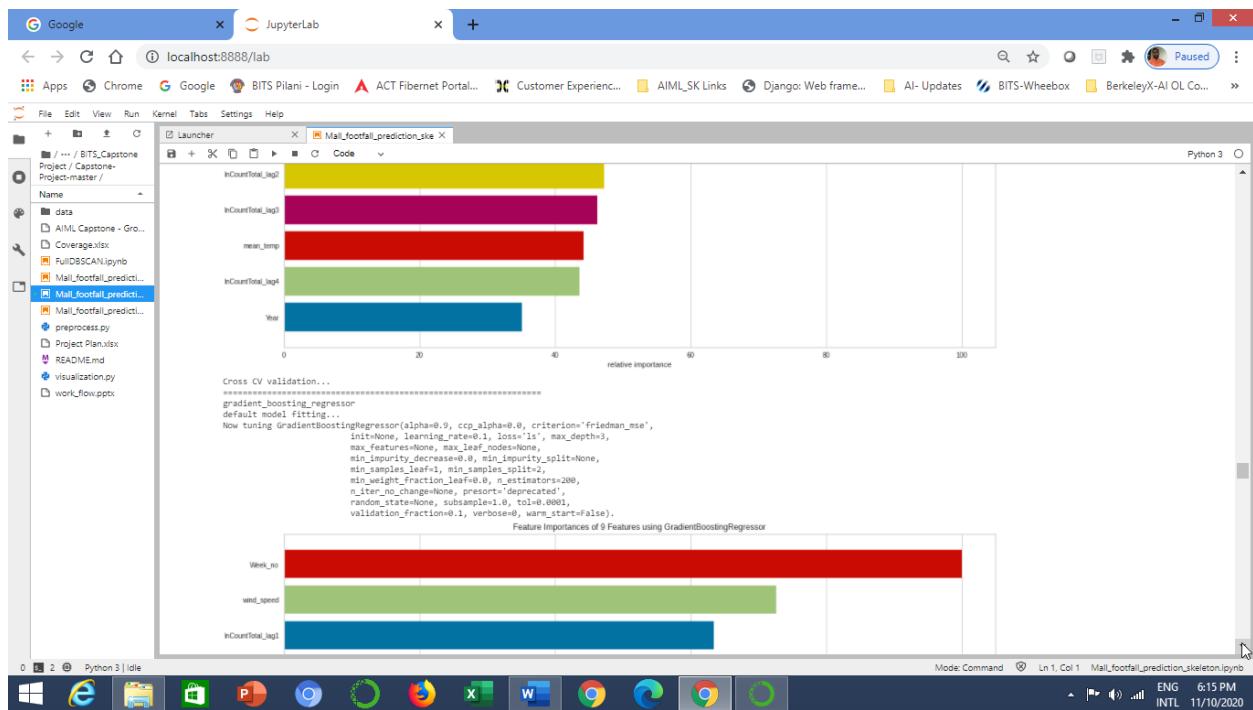
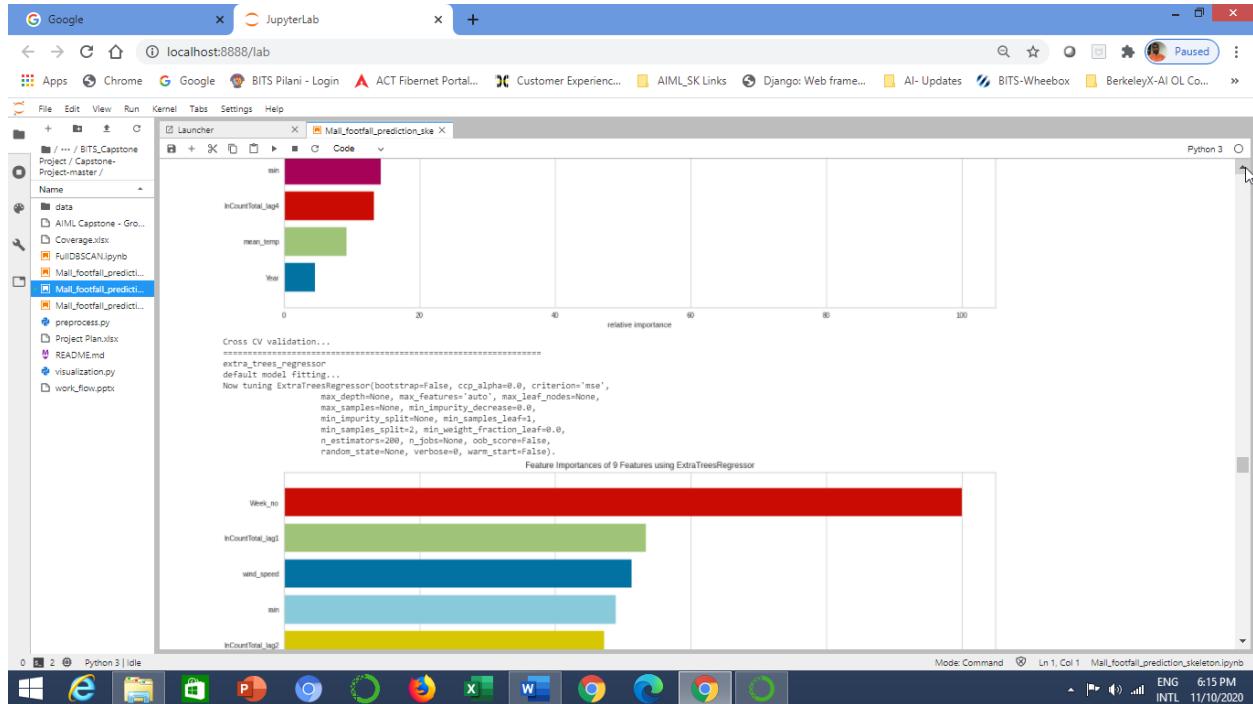
work\_flow.pptx

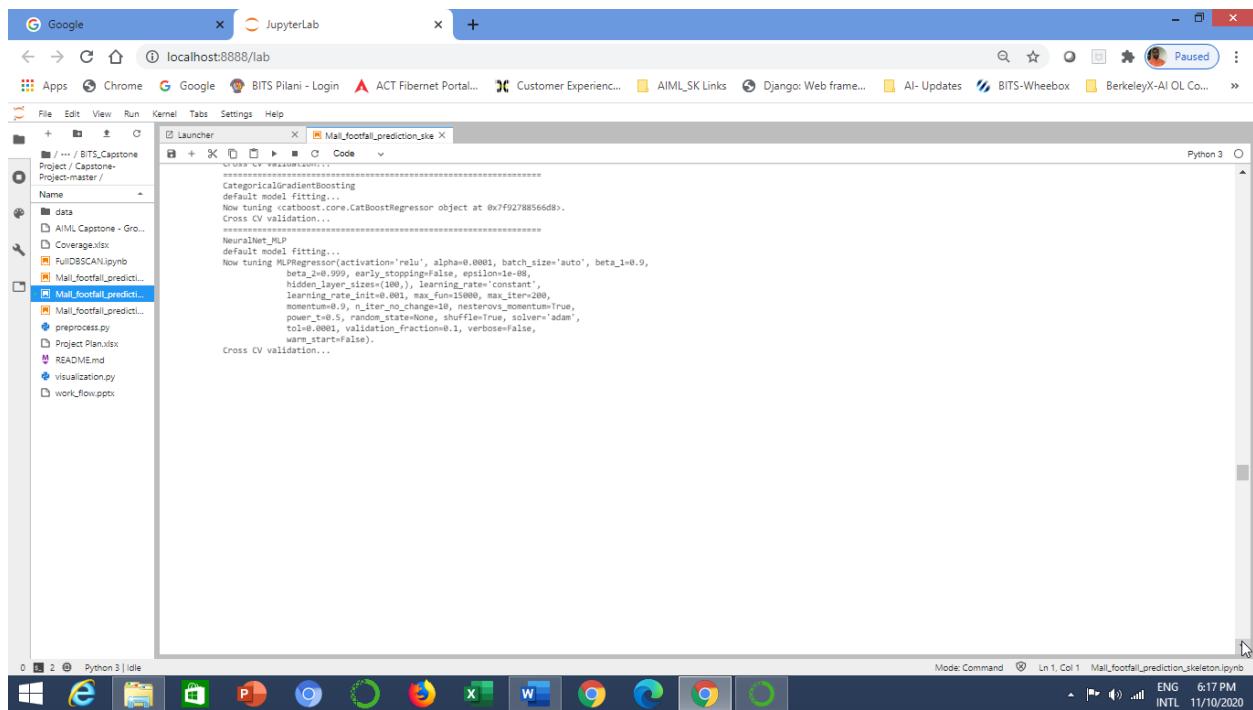
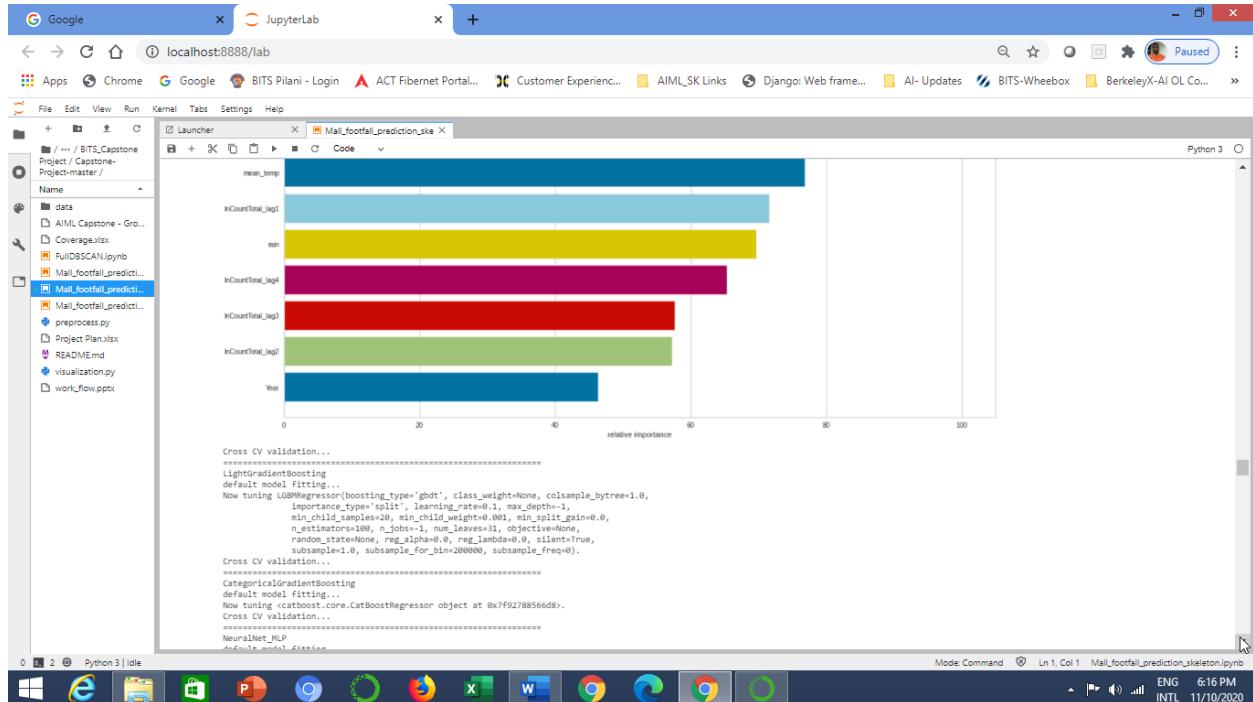
```
Cross CV validation...
=====
default model fitting...
Now tuning BaggingRegressor(base_estimator=None, bootstrap=True, bootstrap_features=False,
   max_features=1.0, max_samples=1.0, n_estimators=200,
   oob_score=False, random_state=None, verbose=0,
   warm_start=False).
Cross CV validation...
=====
random_forest
default model fitting...
Now tuning RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
   max_depth=None, max_features='auto', max_leaf_nodes=None,
   max_samples=None, min_impurity_decrease=0.0,
   min_impurity_split=None, min_samples_leaf=1,
   min_samples_split=2, min_weight_fraction_leaf=0.0,
   n_estimators=200, n_jobs=None, oob_score=False,
   random_state=None, verbose=0, warm_start=False).
Feature Importances of 9 Features using RandomForestRegressor
```

Feature	Importance
Wind_00	High (Red)
wind_speed	Medium (Green)
inCountTotal_leap1	Low (Blue)
inCountTotal_leap2	Low (Light Blue)
inCountTotal_leap3	Low (Yellow)
sun	Low (Purple)
inCountTotal_leap4	Low (Red)

0 2 Python 3 | idle Mode: Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

11/10/2020 6:14 PM





Google JupyterLab localhost:8888/lab

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Project / Capstone-Project-master /

Name

- data
- AIML Capstone - Gro...
- Coverage.xlsx
- FullDBSCAN.ipynb
- Mail\_footfall\_predict...**
- Mail\_footfall\_predict...
- preprocess.py
- Project Plan.xlsx
- README.md
- visualisation.py
- work\_flow.pptx

Launcher Mail\_footfall\_prediction\_ske

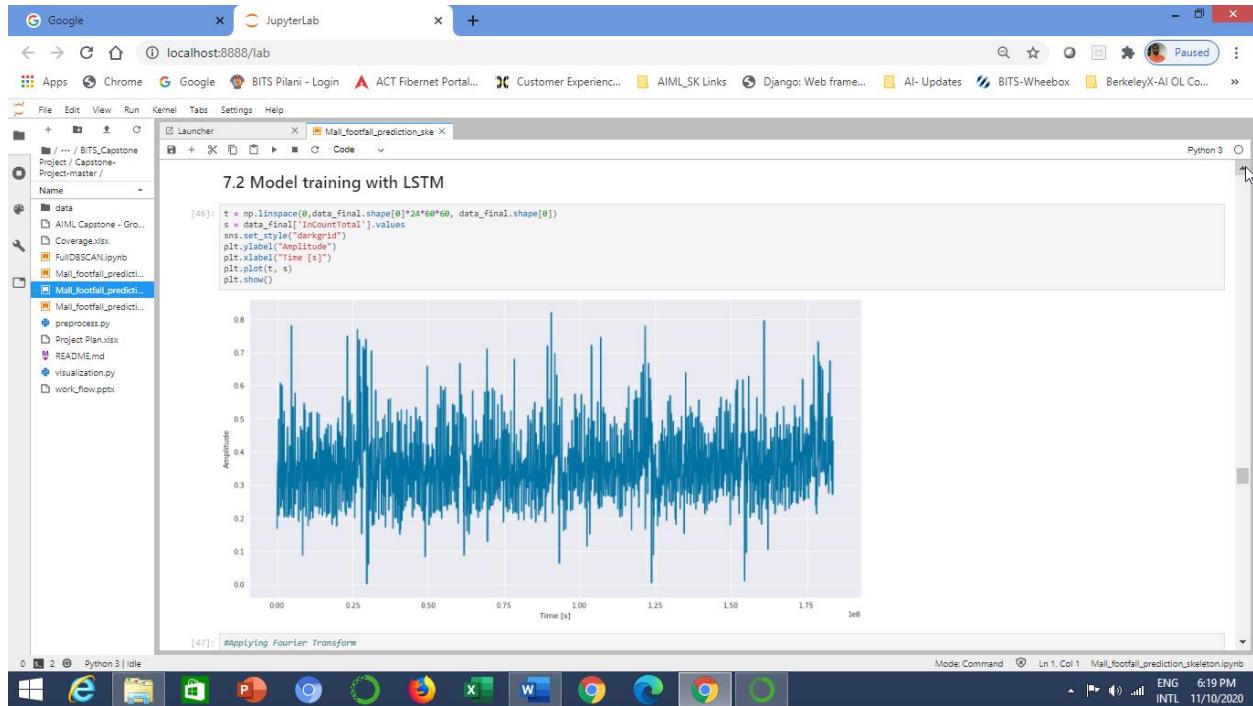
```
[44]: loss_weekly_df.sort_values('mape').head(15)
```

	Model_type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
16	passive_aggressor_weekly_data_rscv_model	7.422109	0.074722	-0.046541	0.204578	-0.008666	0.273353	0.386368	-0.051940	-44.849591	-32.660832
10	elastic_net_weekly_data_rscv_model	7.004383	0.088680	-0.095026	0.216313	-0.024168	0.294754	0.286523	-0.223103	-41.080706	-28.891947
7	ridge_regression_weekly_data_rscv_model	7.716040	0.087055	-0.082116	0.218258	-0.019288	0.295067	0.285005	-0.223705	-41.027590	-28.838831
18	sgd_weekly_data_default_model	7.726780	0.079596	-0.035988	0.210905	0.000700	0.282128	0.346399	-0.120583	-43.269722	-31.080943
0	linear_regression_weekly_data_default_model	7.728431	0.086551	-0.079071	0.210259	-0.018129	0.294107	0.289392	-0.210233	-41.180393	-28.991635
42	extra_trees_regressor_weekly_data_default_model	7.731591	0.088191	-0.066664	0.217247	-0.014674	0.296979	0.277597	-0.241559	-40.706287	-28.817559
1	linear_regression_weekly_data_rscv_model	7.813878	0.095665	-0.198211	0.223809	-0.033625	0.309299	0.214370	-0.346795	-38.672300	-26.483541
12	huberRegressor_weekly_data_default_model	7.971273	0.099953	-0.109871	0.227959	-0.029165	0.316154	0.179162	-0.407195	-37.576330	-29.387572
4	lasso_regressor_weekly_data_rscv_model	7.988315	0.102159	-0.139113	0.229825	-0.036573	0.319653	0.161045	-0.438202	-37.030541	-28.441783
13	huberRegressor_weekly_data_rscv_model	8.028735	0.105962	-0.172223	0.229333	-0.028012	0.317320	0.173905	-0.417551	-37.392231	-29.520473
19	rgd_weekly_data_rscv_model	8.085729	0.092049	-0.116543	0.230060	-0.032379	0.303995	0.244075	-0.295871	-39.639116	-27.447158
55	CategoricalBoostingBooster_weekly_data_rscv_model	8.115489	0.102521	-0.121315	0.223487	-0.034995	0.332041	0.094594	-0.552125	-35.124685	-22.936106
58	NeuralNetMLP_weekly_data_rscv_model	8.129184	0.092107	-0.172029	0.229737	-0.037574	0.303041	0.248600	-0.264668	-39.620193	-27.412435
34	adaBoost_weekly_data_rscv_model	8.185287	0.104819	-0.114693	0.234424	-0.031501	0.323757	0.139207	-0.475646	-36.388095	-24.199337
31	svm_regressor_weekly_data_rscv_model	8.305405	0.104254	-0.110780	0.235685	-0.029748	0.322886	0.143840	-0.467703	-38.523201	-24.334262

```
[45]: loss_weekly_df.sort_values('mape').to_csv('weekly_data_model_loss_values.csv')
```

## 7.2 Model training with LSTM

```
[46]: t = np.linspace(0,data_final.shape[0]*24*60*60, data_final.shape[0])
s = data_final['InCountTotal'].values
sns.set(style="darkgrid")
plt.ylabel("Amplitude")
plt.xlabel("Time [s]")
plt.plot(t, s)
plt.show()
```



Google JupyterLab

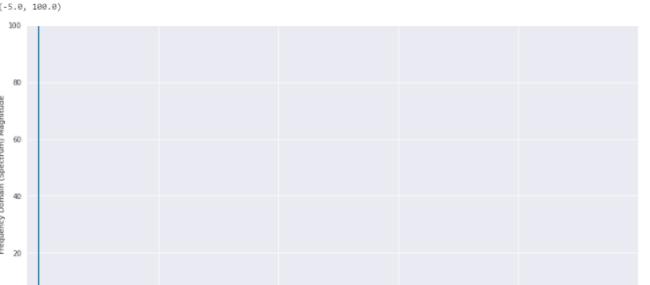
localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_skeleton.ipynb

[47]: #Applying Fourier Transform  
fft = fftpack.fft(s[0:365])  
  
f\_s = 1000  
freqs = fftpack.fftfreq(len(s[0:365])) \* f\_s  
fig, ax = plt.subplots()  
  
ax.stem(freqs, np.abs(fft))  
ax.set\_xlabel('Frequency in Hertz [Hz]')  
ax.set\_ylabel('Frequency Domain (Spectrum) Magnitude')  
ax.set\_xlim(10, f\_s / 2)  
ax.set\_ylim(-5, 100)

[47]: (-5, 0, 100.0)



Python 3 | idle

Mode Command

Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

ENG 6:19 PM  
INTL 11/10/2020

Google JupyterLab

localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

[48]:

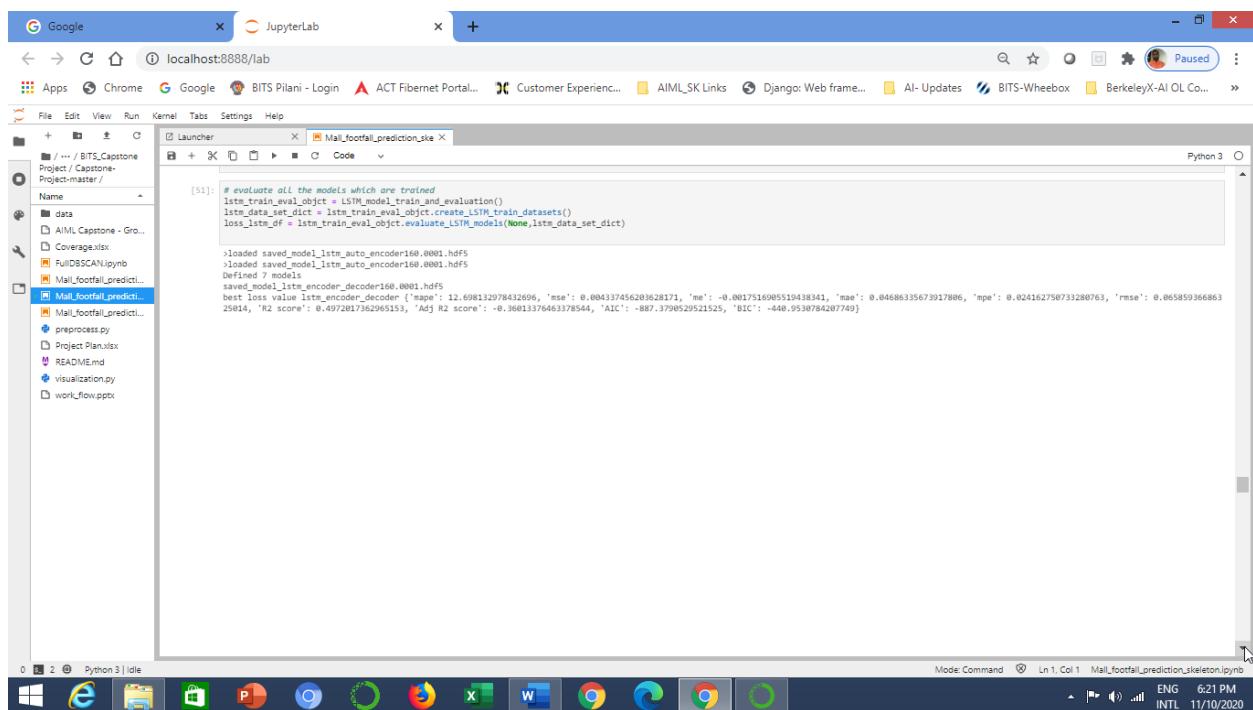
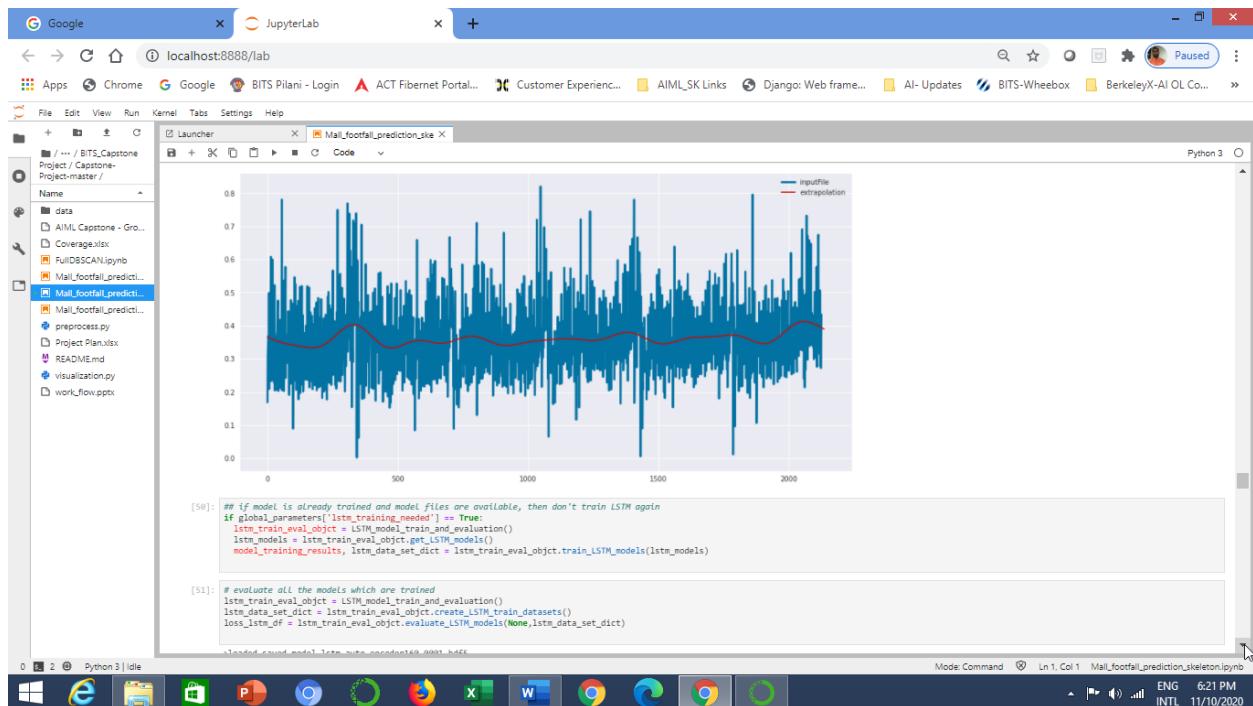
```
def fourierExtrapolation(x, n_predict):
    n = x.size
    n_harm = 10
    t = np.arange(0, n)           # number of harmonics in model
    p = np.polyfit(t, x, 1)       # find linear trend in x
    x_notrend = x - p[0] * t     # detrended x
    x_frequencies = ftfpack.fft(x_notrend) # detrended x in frequency domain
    f = ftfpack.fftfreq(n)        # frequencies
    indexes = list(range(n))     # sort indexes by frequency, Lower > higher
    indexes.sort(key = lambda i: np.abs(f[i]))
    t = np.arange(0, n + n_predict)
    restored_sig = np.zeros(t.size)
    for i in indexes[1:n_harm+1]:
        ampli = np.absolute(x_frequencies[i]) / n # amplitude
        phase = np.angle(x_frequencies[i]) # phase
        restored_sig += ampli * np.cos(2 * np.pi * f[i] * t + phase)
    return restored_sig + p[0] * t

x = data_final['IncountTotal'].values
extrapolation = fourierExtrapolation(x, 10)
plt.plot(np.arange(0, x.size), x, label = 'Inputfile', linewidth = 3)
plt.plot(np.arange(0, extrapolation.size), extrapolation, 'r', label = 'extrapolation')
plt.legend()
```

[48]:

Frequency in Hertz [Hz]

Python 3



```
saved_model_lstm_simple_model160.0001.hdf5
best loss value lstm_simple_model {'mape': 15.81209562147549, 'mse': 0.006353162963685339, 'me': 0.0005099347531410092, 'mae': 0.057859479715478355, 'mpe': 0.03706400852564353, 'rmse': 0.079706730980547
3, 'R2 score': 0.2653468504058893, 'Adj R2 score': -0.9922164174919772, 'AIC': -866.4661269441533, 'BIC': -360.0481524127757}

saved_model_cnn_lstm_encoder_decoder160.0001.hdf5
best loss value cnn_lstm_encoder_decoder {'mape': 13.628688667048838, 'mse': 0.0050473882485827975, 'me': -0.0072889855293858, 'mae': 0.05097148072611497, 'mpe': 0.00814313436095705, 'rmse': 0.07104497
342235268, 'R2 score': 0.41490635790120886, 'Adj R2 score': -0.5827533138826282, 'AIC': -855.2434817709654, 'BIC': -498.8175072395878}
```

The screenshot shows a JupyterLab environment running in a browser window. The title bar indicates the URL is `localhost:8888/lab`. The left sidebar displays a file tree for a project named "BITS\_Capstone". The current file, `Mail_footfall_prediction_skeleton.ipynb`, is selected. A code cell in the main area contains the following Python code and its output:

```
saved_model_lstm_encoder_decoder_outlier_removed168_001.hdf5
best loss value lstm_encoder_decoder_outlier_removed {'mape': 12.764863948547485, 'mse': 0.00389735335314146084, 'me': -0.0047669233170767603, 'mae': 0.0476011305715913, 'mpe': 0.0106138774173531, 'rmse': 0.05242879625613815, 'R2 score': 0.558912107426477, 'Adj R2 score': -0.23499145707726887, 'AIC': -898.956925459551, 'BIC': -453.8807905063174}
```

The screenshot shows a Jupyter Notebook interface with several tabs open. The active tab is titled 'Mail\_footfall\_prediction.ipynb'. The code cell contains the following Python code:

```
saved_model_lstm_simple_model_outlier_removed_0001.hdf5
best loss value lstm_simple_model_outlier_removed ['nape': 15.478077555342031, 'msa': -0.006207543766680563, 'me': -0.00525956851250195, 'mee': 0.05796593861435797, 'mpe': 0.025893722402587225, 'rmse': 0.878796714397884, 'R2 score': 0.28471146812047404, 'Adj R2 score': -0.9678434866681442, 'AIC': -801.217897898874, 'BIC': -356.0525963112641]
```

The notebook also lists other files in the directory on the left, including 'saved\_model\_lstm\_auto\_encoder\_MLP160\_0001.hdf5' and 'best loss value lstm\_auto\_encoder\_MLP ['nape': 14.49881529061747, 'msa': 0.005269380528961401, 'me': -0.004683848524320185, 'mee': 0.05442220481916744, 'mpe': 0.023513101843763855, 'rmse': 0.07258994785065986, 'R2 score': 0.392824114414221, 'Adj R2 score': -0.6697334185360893, 'AIC': -835.6301069599342, 'BIC': -390.4648053745109].

```
[tf.Tensor 'past_inputs_3:0' shape=(None, 7, 19) dtype=float32], [tf.Tensor 'future_input_1:0' shape=(None, 1, 8) dtype=float32], [

```
loaded saved_model_lstm_encoder_decoder_160_0001.hdfs
loaded saved_model_cnn_lstm_encoder_decoder_outlier_removed160_0001.hdfs
>loaded saved_model_cnn_lstm_encoder_decoder_outlier_removed160_0001.hdfs
Loaded 4 models
```


```

```
loaded saved_model_lstm_encoder_decoder_outlier_removed160_0001.hdfs
loaded saved_model_cnn_lstm_encoder_decoder_outlier_removed160_0001.hdfs
Loaded 2 models
```

```
>loaded saved_model_lstm_encoder_decoder_outlier_removed160_0001.hdfs
>loaded saved_model_cnn_lstm_encoder_decoder_outlier_removed160_0001.hdfs
>loaded saved_model_lstm_auto_encoder_MLP160_0001.hdfs
Loaded 3 models
(None, 1, 1)
```

The screenshot shows a JupyterLab interface with a code cell containing Python code. The output of the code shows the loading of multiple saved models from an HDFS path. The code is as follows:

```
>loaded saved_model_lstm_encoder_decoder160_0001.hdfs
>loaded saved_model_cnn_lstm_encoder_decoder160_0001.hdfs
>loaded saved_model_lstm_encoder_decoder_outlier_removed160_001.hdfs
>loaded saved_model_cnn_lstm_encoder_decoder_outlier_removed160_001.hdfs
>loaded saved_model_lstm_auto_encoder_MLP160_0001.hdfs
Loaded 4 models
(None, 1, 1)

>loaded saved_model_stacked_ensemble_lstm116.hdfs
>loaded saved_model_stacked_ensemble_lstm116.hdfs
```

The status bar at the bottom indicates "Mode: Command" and "Ln 1, Col 1". The system tray shows "ENG 6:25 PM INTL 11/10/2020".

The screenshot shows a JupyterLab interface with a code cell containing Python code. The output of the code shows the loading of multiple saved models from an HDFS path. The code is as follows:

```
>loaded saved_model_stacked_ensemble_lstm116.hdfs
>loaded saved_model_stacked_ensemble_lstm116.hdfs
>loaded saved_model_stacked_ensemble_lstm116.hdfs
>loaded saved_model_stacked_ensemble_lstm116.hdfs
Loaded 4 models
```

The status bar at the bottom indicates "Mode: Command" and "Ln 1, Col 1". The system tray shows "ENG 6:26 PM INTL 11/10/2020".

The screenshot shows a Jupyter Notebook interface with the following details:

- Header:** Google, JupyterLab, localhost:8888/lab
- Toolbar:** Apps, Chrome, Google, BITS Pilani - Login, ACT Fibernet Portal..., Customer Experience..., AIML\_SK Links, Django: Web frame..., AI- Updates, BITS-Wheebox, BerkeleyX-AI OL Co... Paused
- File Menu:** File, Edit, View, Run, Kernel, Tabs, Settings, Help
- Launcher:** Mail\_footfall\_prediction\_skeleton
- Code Cell 1:** Loaded saved\_model\_stacked\_ensemble\_lstm116.hdf5  
Loaded saved\_model\_stacked\_ensemble\_lstm216.hdf5  
Loaded 2 models
- Code Cell 2:** [52]:

```
loss_values.all_models.append(loss_lstm_df)
loss_lstm_df.sort_values("mape").to_csv("lstm_best_models.csv")
loss_lstm_df.sort_values("mape").head(25)
```
- Output Headers:** Model\_type, mape, mse, me, mae, rmse, R2 score, Adj R2 score, AIC, BIC
- Bottom Bar:** Mode Command, Ln 1, Col 1, Mail\_footfall\_prediction\_skeleton.ipynb, Python 3 | idle
- Taskbar:** Windows, Edge, File Explorer, File, Firefox, X, W, Chrome, Edge, Google Chrome, Microsoft Edge
- System Status:** ENG, 6:26 PM, INTL, 11/10/2020

Google JupyterLab

localhost:8888/lab

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Launcher Mail\_footfall\_prediction\_skeleton.py

[52]: loss\_values\_all\_models.append(loss\_lstm\_df)  
loss\_lstm\_df.sort\_values('mape').to\_csv("lstm\_best\_models.csv")  
loss\_lstm\_df.sort\_values('mape').head(25)

[52]:

	Model_Type	mape	mse	me	mae	mpe	rmse	R2 score	Adj R2 score	AIC	BIC
16	ensemble_of_ensembles_lstm3	10.891565	0.003314	-0.008190	0.041015	0.000334	0.057587	0.618135	-0.050130	-93.017665	-487.832386
11	stacked_ensemble_lstm2	10.891049	0.003273	-0.009190	0.040578	0.011875	0.057212	0.628832	-0.037213	-93.016757	-490.451456
10	stacked_ensemble_lstm1	10.902380	0.003348	-0.007735	0.040937	0.002332	0.057858	0.614270	-0.060757	-93.019348	-485.737847
8	ensemble_lstm_outlier_removed	10.897030	0.003331	-0.007373	0.041243	0.003090	0.057713	0.616199	-0.055453	-93.01958811	-486.790509
12	stacked_ensemble_lstm3	10.902780	0.003274	-0.008725	0.040974	0.010284	0.057217	0.622765	-0.037395	-93.0179831	-491.414529
15	ensemble_of_ensembles_lstm2	11.003165	0.003297	-0.006856	0.041194	0.016453	0.057417	0.620133	-0.044653	-93.0113825	-488.948524
14	ensemble_of_ensembles_lstm1	11.036401	0.003289	-0.005891	0.041269	0.003800	0.057345	0.612166	-0.042069	-93.0153688	-487.470567
13	stacked_ensemble_lstm3	11.099465	0.003300	-0.000118	0.041117	0.019013	0.057443	0.619783	-0.045596	-93.026211	-488.760910
9	ensemble_lstm_with_auto_encoder	11.441193	0.003490	-0.004543	0.042928	0.009848	0.059078	0.597824	-0.105983	-92.2135239	-476.959991
4	cnn_lstm_encoder_decoder_outlier_removed	11.704837	0.003955	-0.009994	0.044963	0.004986	0.062885	0.544323	-0.253112	-99.072025	-450.741903
7	ensemble_lstm	12.305297	0.004262	-0.004516	0.045603	0.016154	0.056253	0.505988	-0.364488	-100.103482	-444.677461
0	lstm_encoder_decoder	12.695133	0.004337	-0.001752	0.048663	0.024163	0.056539	0.497202	-0.360134	-887.790533	-440.953078
3	lstm_encoder_decoder_outlier_removed	12.764844	0.003897	-0.004757	0.047060	0.011061	0.062429	0.550912	-0.234991	-98.960093	-453.800791
2	cnn_lstm_encoder_decoder	13.028689	0.005047	0.00971	0.008144	0.071045	0.0414906	-0.082753	-855.243486	-408.817507	
6	lstm_auto_encoder_lstm	14.998823	0.005269	-0.004605	0.04422	0.023518	0.072590	0.392824	-0.499733	-855.530107	-500.464805
5	lstm_impute_model_outlier_removed	15.470578	0.006208	-0.002533	0.057966	0.025904	0.078788	0.284711	-0.67043	-201.217898	-350.022596
1	lstm_impute_model	15.813906	0.006353	0.001661	0.057859	0.037064	0.079707	0.263541	-0.992216	-804.496127	-380.040152

[53]: h2o\_model\_obj = h2o.model(data\_final,dataset\_type='Normal')  
loss\_h2o\_model = h2o\_model\_obj.train\_and\_validate\_models()  
loss\_values\_all\_models.append(loss\_h2o\_model)

Mode Command Ln 1 Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3 | die

ENG INTEL 11/10/2022

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Launcher Mail\_footfall\_prediction\_ske

Python 3

## 7.3 Model training with H2O

```
[53]: h2o_model_obj = h2o_model(data_final,dataset_type='Normal')
loss_h2o_model = h2o_model_obj.train_and_evaluate_models()
loss_values_all_models.append(loss_h2o_model)

Defined 2 models
Parse progress: [██████████] 100%
Parse progress: [██████████] 100%
gbm Model Build progress: [██████████] 100%
ModelMetricsRegression: gbm
** Reported on test data. **

MSE: 0.0064561708328316385
RMSE: 0.080859807637998
MAE: 0.05299778405752884
MMAE: 0.0515146145788769
Mean Residual Deviance: 0.0064561708328316385

gbm prediction progress: [██████████] 100%
H2OGradientBoosting
default H2O model fitting...
Model Details
=====
H2OGradientBoostingEstimator : Gradient Boosting Machine
Model Key: GBM_Model_python_1604675476624_1

Model Summary:
  number_of_trees    number_of_internal_trees   model_size_in_bytes   min_depth   max_depth   mean_depth   min_leaves   max_leaves   mean_leaves
  0                  100.0                      100.0                35714.0      5.0        5.0        5.0        13.0       32.0       23.74

ModelMetricsRegression: gbm
** Reported on train data. **
MSE: 0.0010644767301113676
```

The screenshot shows a JupyterLab environment with multiple tabs open. The active tab is titled 'Mail\_footfall\_prediction\_skeleton.ipynb'. The code cell contains Python code for model evaluation, including metrics for training, validation, and cross-validation. The results are displayed as tables.

```
ModelMetricsRegression: gbm
** Reported on train data. **

MSE: 0.001041673915113676
RMSE: 0.03262517861778938
MAE: 0.0242208401952553
RMSLE: 0.02387448577107221
Mean Residual Deviance: 0.001041673915113676

ModelMetricsRegression: gbm
** Reported on validation data. **

MSE: 0.004510840000056501
RMSE: 0.0673537972222472
MAE: 0.04699163860157243
RMSLE: 0.049134513348205536
Mean Residual Deviance: 0.004510840000056501

ModelMetricsRegression: gbm
** Reported on cross-validation data. **

MSE: 0.065319662810018405
RMSE: 0.27233918260547785
MAE: 0.053787266102563
RMSLE: 0.05297893747804434
Mean Residual Deviance: 0.065319662810018405

Cross-Validation Metrics Summary:
```

	mean	sd	cv_1_valid	cv_2_valid	cv_3_valid	cv_4_valid	cv_5_valid	
0	mse	0.0503650583	0.002840793	0.049087662	0.05490401	0.051075403	0.04928869	0.04746964
1	mean_residual_deviance	0.0553178864	5.017278E-4	0.0555079474	0.006116455	0.004955005	0.0049431133	0.005056911
2	mse	0.0553178864	5.017278E-4	0.0555079474	0.006116455	0.004955005	0.0049431133	0.005056911
3	r2	0.42012027	0.04499243	0.42245987	0.734974	0.475468	0.45186938	0.3735207
4	residual_deviance	0.0553178864	5.017278E-4	0.0555079474	0.006116455	0.004955005	0.0049431133	0.005056911
5	rmse	0.07236107	0.003382113	0.074215548	0.07820777	0.07046279	0.07030728	0.07111196
6	rmsle	0.052912988	0.0028455098	0.053466495	0.057607107	0.050758954	0.052099895	0.050633244

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Python 3

Scoring History:

	timestamp	duration	number_of_trees	training_rmse	training_mae	training_deviance	validation_rmse	validation_mae	validation_deviance
0	2020-11-06 16:12:14	10.465 sec	0.0	0.069067	0.069117	0.009229	0.091773	0.065104	0.008422
1	2020-11-06 16:12:14	10.489 sec	1.0	0.091638	0.068047	0.008398	0.087523	0.062152	0.007680
2	2020-11-06 16:12:14	10.516 sec	2.0	0.087845	0.063873	0.007717	0.084178	0.059921	0.007085
3	2020-11-06 16:12:14	10.531 sec	3.0	0.084651	0.061342	0.007145	0.081511	0.058012	0.006644
4	2020-11-06 16:12:14	10.546 sec	4.0	0.081801	0.059300	0.006891	0.079231	0.056236	0.006278
5	2020-11-06 16:12:14	10.563 sec	5.0	0.079309	0.057856	0.006390	0.077256	0.054840	0.005969
6	2020-11-06 16:12:14	10.580 sec	6.0	0.077145	0.055863	0.005951	0.075467	0.053355	0.005695
7	2020-11-06 16:12:14	10.597 sec	7.0	0.075059	0.054277	0.005635	0.073827	0.052209	0.005450
8	2020-11-06 16:12:14	10.614 sec	8.0	0.073190	0.052754	0.005357	0.072417	0.051166	0.005244
9	2020-11-06 16:12:14	10.630 sec	9.0	0.071613	0.051421	0.005128	0.071593	0.050347	0.005105
10	2020-11-06 16:12:14	10.645 sec	10.0	0.070323	0.050410	0.004945	0.070658	0.049710	0.004993
11	2020-11-06 16:12:14	10.660 sec	11.0	0.068827	0.049244	0.004737	0.069758	0.049027	0.004866
12	2020-11-06 16:12:14	10.678 sec	12.0	0.067552	0.048236	0.004563	0.069144	0.048511	0.004781
13	2020-11-06 16:12:14	10.716 sec	13.0	0.066140	0.047239	0.004575	0.068346	0.047766	0.004699
14	2020-11-06 16:12:14	10.730 sec	14.0	0.064974	0.046415	0.004222	0.068607	0.047599	0.004707
15	2020-11-06 16:12:14	10.743 sec	15.0	0.063799	0.045591	0.004070	0.068228	0.047252	0.004555
16	2020-11-06 16:12:14	10.758 sec	16.0	0.062785	0.044888	0.003942	0.067906	0.047007	0.004611
17	2020-11-06 16:12:14	10.774 sec	17.0	0.061707	0.044226	0.003808	0.067771	0.046937	0.004593
18	2020-11-06 16:12:14	10.794 sec	18.0	0.060552	0.043525	0.003663	0.067279	0.046599	0.004526
19	2020-11-06 16:12:14	10.811 sec	19.0	0.059552	0.042773	0.003546	0.067425	0.046591	0.004546

See the whole table with table.as\_data\_frame()

Variable Importances:

variable	relative_importance	scaled_importance	percentage
Weekday	14.654948	1.000000	0.741544

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 idle

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Python 3

Var180181 Importances:

variable	relative_importance	scaled_importance	percentage
Weekday	14.654948	1.000000	0.241344
Weekend	8.004540	0.607199	0.145544
InCountTotal_lag1	6.570343	0.448030	0.108129
Day	6.081395	0.414609	0.100083
InCountTotal_lag5	4.907845	0.334651	0.080756
InCountTotal_lag3	3.485950	0.237686	0.057364
mean_temp	2.9177407	0.198937	0.048012
InCountTotal_lag7	2.802591	0.191108	0.048123
InCountTotal_lag6	2.665076	0.181731	0.043690
InCountTotal_lag2	2.391912	0.160376	0.038706
InCountTotal_lag4	1.372151	0.093567	0.022582
wind_speed	1.174959	0.080120	0.019337
Month	0.983315	0.067052	0.016183
School_holidays	0.761623	0.051935	0.012534
rain	0.514946	0.035114	0.008475
University_holidays	0.434950	0.029659	0.007158
UKBankHoliday	0.148279	0.010111	0.002440
high_wind	0.011723	0.000799	0.001193
low_temp	0.0098832	0.000602	0.001145
high_temp	0.0001965	0.000099	0.000022

Now tuning...

gbm Grid Build progress: [██████████] 100%

col_sample_rate	... model_ids	mse
0.9	H2GradientBoosting_model_16	0.005276268724219805
0.4	H2GradientBoosting_model_8	0.005354788695899225
0.8	H2GradientBoosting_model_10	0.005355325283598369

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 idle

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Launcher Mail\_footfall\_prediction\_ske

Python 3

Var180181 Importances:

variable	relative_importance	scaled_importance	percentage
Weekday	14.654948	1.000000	0.241344
Weekend	8.004540	0.607199	0.145544
InCountTotal_lag1	6.570343	0.448030	0.108129
Day	6.081395	0.414609	0.100083
InCountTotal_lag5	4.907845	0.334651	0.080756
InCountTotal_lag3	3.485950	0.237686	0.057364
mean_temp	2.9177407	0.198937	0.048012
InCountTotal_lag7	2.802591	0.191108	0.048123
InCountTotal_lag6	2.665076	0.181731	0.043690
InCountTotal_lag2	2.391912	0.160376	0.038706
InCountTotal_lag4	1.372151	0.093567	0.022582
wind_speed	1.174959	0.080120	0.019337
Month	0.983315	0.067052	0.016183
School_holidays	0.761623	0.051935	0.012534
rain	0.514946	0.035114	0.008475
University_holidays	0.434950	0.029659	0.007158
UKBankHoliday	0.148279	0.010111	0.002440
high_wind	0.011723	0.000799	0.001193
low_temp	0.0098832	0.000602	0.001145
high_temp	0.0001965	0.000099	0.000022

Now tuning...

gbm Grid Build progress: [██████████] 100%

col_sample_rate	... model_ids	mse
0.9	H2GradientBoosting_model_16	0.005276268724219805
0.4	H2GradientBoosting_model_8	0.005354788695899225
0.8	H2GradientBoosting_model_10	0.005355325283598369

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 idle

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Launcher Mail\_footfall\_prediction\_ske

```
See the whole table with table.as_data_frame()
Now tuning
gbm Grid Build progress: [██████████] 100%
col_sample_rate ... model_ids mse
0 0.9 ... H2GradientBoosting_model_16 0.005276269782419805
1 0.4 ... H2GradientBoosting_model_8 0.00535478869589925
2 0.8 ... H2GradientBoosting_model_10 0.005355326283598169
3 0.6 ... H2GradientBoosting_model_12 0.005355326283598169
4 0.30000000000000004 ... H2GradientBoosting_model_18 0.00548433942931528
5 0.8 ... H2GradientBoosting_model_20 0.005488842242320938
6 0.9 ... H2GradientBoosting_model_1 0.0084367659533478
7 0.6000000000000001 ... H2GradientBoosting_model_2 0.00851072258204683
8 0.9 ... H2GradientBoosting_model_4 0.009510746288620398
9 0.5 ... H2GradientBoosting_model_13 0.009518276470232682
10 0.1 ... H2GradientBoosting_model_14 0.0095567794268239815
11 0.2 ... H2GradientBoosting_model_9 0.00963333316098286
12 0.2 ... H2GradientBoosting_model_15 0.00963333316098286
13 0.5 ... H2GradientBoosting_model_14 0.00981940923044877365
14 0.7000000000000001 ... H2GradientBoosting_model_12 0.009815888827535854
15 0.5 ... H2GradientBoosting_model_7 0.009819801128557859
16 0.4 ... H2GradientBoosting_model_17 0.00984333316098286
17 0.5 ... H2GradientBoosting_model_1 0.00984333316098286
18 0.5 ... H2GradientBoosting_model_15 0.0098725151208345135
19 1.0 ... H2GradientBoosting_model_3 0.0097400000074664849
```

[28 rows x 7 columns]

ModelMetricsRegression: gbm  
\*\* Reported on test data. \*\*

MSE: 0.0061380121217705783  
RMSE: 0.078136710723686  
MAE: 0.052256236858579  
MSLE: 0.05136424098546101  
Mean Residual Deviance: 0.0061380121217705783

gbm prediction progress: [██████████] 100%
gbm Grid Build progress: [██████████] 100%
stackedensemble prediction progress: [██████████] 100%
drf Model Build progress: [██████████] 100%

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyb

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File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```
executeonnewprediction progress: [██████████] 100%
drf Model Build progress: [██████████] 100%
```

ModeMetricsRegression: drf  
\*\* Reported on test data. \*\*

MSE: 0.005907311609567155  
RMSE: 0.07480916789936176  
MAE: 0.049854720317412  
MSLE: 0.0547787203843846  
Mean Residual Deviance: 0.005907311609567155

drf prediction progress: [██████████] 100%
H2RandomForest
Default H2o model fitting...
Model Details
ensemble
H2RandomForestEstimator : Distributed Random Forest
Model Key: DRF\_model\_python\_1684675476624\_2

Model Summary:

	number_of_trees	number_of_internal_trees	model_size_in_bytes	min_depth	max_depth	mean_depth	min_leaves	max_leaves	mean_leaves
0	100.0	100.0	10183534.0	20.0	20.0	20.0	616.0	904.0	8064.2

ModelMetricsRegression: drf  
\*\* Reported on train data. \*\*

MSE: 0.005429523883322289  
RMSE: 0.0736853838347456  
MAE: 0.05895753504561063  
MSLE: 0.055901393291128435  
Mean Residual Deviance: 0.005429523883322289

ModeMetricsRegression: drf  
\*\* Reported on validation data. \*\*

MSE: 0.0043961185489001154  
RMSE: 0.06639323151175328  
MAE: 0.0449835304119999  
MSLE: 0.047040989895944439

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyb

Google JupyterLab localhost:8888/lab

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Launcher Mail\_footfall\_prediction\_ske

RMSLE: 0.047794098895946439  
Mean Residual Deviance: 0.0043961185808981154

Model: ReportricsRegression: drf  
\*\* Reported on cross-validation data. \*\*

MSE: 0.005519967435013022  
RMSE: 0.0742094832878924  
MAE: 0.051028532325784223  
RMSLE: 0.053912922537841919  
Mean Residual Deviance: 0.005519967435013022

Cross-Validation Metrics Summary:

	mean	sd	cv_1_valid	cv_2_valid	cv_3_valid	cv_4_valid	cv_5_valid	
0	mae	0.051014718	0.002527442	0.049250312	0.054646354	0.051607933	0.051496435	0.048070554
1	mean_residual_deviance	0.005622096	3.8102508E-4	0.0051807966	0.0061551863	0.0053774499	0.0057435956	0.0053226175
2	mse	0.005522094	3.8102508E-4	0.0051807966	0.0061551863	0.0053774499	0.0057435956	0.0053226175
3	r2	0.96991028	0.044567473	0.45676354	0.37153277	0.43188974	0.38186738	0.34249803
4	residual_deviance	0.0055222092	3.8102508E-4	0.0051807966	0.0061551863	0.0053774499	0.0057435956	0.0053226175
5	rmse	0.07427642	0.0053262905	0.07197775	0.07845499	0.07338143	0.074661635	0.07295627
6	rsme	0.053871922	0.0028142475	0.05187397	0.05816287	0.052439744	0.0525277002	0.051784242

Scoring History:

	timestamp	duration	number_of_trees	training_rmse	training_mae	training_deviance	validation_rmse	validation_mae	validation_deviance
0	2020-11-06 16:14:29	18.887 sec	0.0	NaN	NaN	NaN	NaN	NaN	NaN
1	2020-11-06 16:14:29	18.925 sec	1.0	0.116327	0.079342	0.015532	0.106379	0.074009	0.011316
2	2020-11-06 16:14:29	18.950 sec	2.0	0.108419	0.074900	0.011755	0.086994	0.064216	0.007568
3	2020-11-06 16:14:29	18.976 sec	3.0	0.101155	0.070498	0.010232	0.079256	0.056435	0.006281
4	2020-11-06 16:14:29	19.002 sec	4.0	0.097672	0.067574	0.009540	0.076359	0.054084	0.005831
5	2020-11-06 16:14:29	19.027 sec	5.0	0.094738	0.065996	0.008975	0.075404	0.052766	0.005686
6	2020-11-06 16:14:29	19.051 sec	6.0	0.091768	0.064870	0.008421	0.073643	0.050779	0.005423
7	2020-11-06 16:14:29	19.079 sec	7.0	0.091434	0.064205	0.008380	0.073055	0.050351	0.005337

The screenshot shows a Jupyter Notebook environment with the following details:

- Header:** Google, JupyterLab, localhost:8888/lab
- Left Sidebar:** Project Capstone, Project:master /, data, Coverage.xlsx, FullDBSCAN.ipynb, Mail\_footfall\_prediction\_ske..., Mail\_footfall\_prediction\_..., preprocess.py, Project Plan.xlsx, README.md, visualization.py, workflow.pptx.
- Code Cell:** Displays a table titled "Mail\_footfall\_prediction\_ske..." with 19 rows of data. The columns are labeled: Date, Time, Duration, Step, InCountTotal\_lag1, InCountTotal\_lag2, InCountTotal\_lag3, InCountTotal\_lag4, InCountTotal\_lag5, InCountTotal\_lag6, InCountTotal\_lag7, Day, mean\_temp, scaled\_importance, importance, and percentage.
- Table Data:** A sample of the data is shown below:

Date	Time	Duration	Step	InCountTotal_lag1	InCountTotal_lag2	InCountTotal_lag3	InCountTotal_lag4	InCountTotal_lag5	InCountTotal_lag6	InCountTotal_lag7	Day	mean_temp	scaled_importance	importance	percentage
2020-11-06	16:14:29	19.079 sec	7.0	0.091434	0.064205	0.008360	0.073059	0.060351	0.005337						
2020-11-06	16:14:29	19.109 sec	8.0	0.089092	0.063287	0.007937	0.072097	0.049451	0.005198						
2020-11-06	16:14:29	19.138 sec	9.0	0.087731	0.061444	0.007997	0.071108	0.049416	0.005056						
2020-11-06	16:14:29	19.177 sec	10.0	0.086591	0.060878	0.007498	0.071937	0.049464	0.005175						
2020-11-06	16:14:29	19.207 sec	11.0	0.085495	0.059841	0.007309	0.071997	0.049496	0.005184						
2020-11-06	16:14:29	19.235 sec	12.0	0.084189	0.058823	0.007088	0.070328	0.048569	0.004946						
2020-11-06	16:14:29	19.258 sec	13.0	0.083504	0.058309	0.006973	0.065569	0.047758	0.004840						
2020-11-06	16:14:29	19.283 sec	14.0	0.082142	0.057337	0.006747	0.070046	0.047885	0.004906						
2020-11-06	16:14:29	19.315 sec	15.0	0.080396	0.056342	0.006459	0.069521	0.047676	0.004833						
2020-11-06	16:14:29	19.358 sec	16.0	0.079610	0.056031	0.006338	0.069847	0.048165	0.004879						
2020-11-06	16:14:29	19.398 sec	17.0	0.078430	0.055269	0.006151	0.069409	0.047952	0.004818						
2020-11-06	16:14:29	19.430 sec	18.0	0.077909	0.054871	0.006070	0.069165	0.047803	0.004784						
2020-11-06	16:14:29	19.472 sec	19.0	0.077423	0.054279	0.005994	0.068504	0.047565	0.004693						

See the whole table with `table.as_data_frame()`

Variable Importances:

variable	relative_importance	scaled_importance	percentage
0 Weekday	147.132009	1.000000	0.174045
1 InCountTotal_lag1	85.596909	0.538693	0.102445
2 Weekn	85.279398	0.574958	0.100886
3 InCountTotal_lag7	74.322266	0.505181	0.087924
4 InCountTotal_lag3	61.646427	0.417782	0.072713
5 InCountTotal_lag5	55.977829	0.380490	0.065222
6 Day	55.946110	0.380275	0.066185
7 InCountTotal_lag6	50.198807	0.341209	0.059386
8 mean_temp	41.174248	0.279688	0.048710
9 InCountTotal_lag2	37.876347	0.257452	0.044808

Google JupyterLab local:8888/lab

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Launcher Mail\_football\_prediction\_ske

	Code
9	InCountTotalJag2 37.876347 0.257452 0.044808
10	InCountTotalJag4 34.828426 0.235375 0.040966
11	Montt 32.841862 0.223231 0.038852
12	wind_speed 29.140993 0.198111 0.034480
13	rain 19.379608 0.131726 0.022926
14	School_holidays 8.744990 0.059441 0.010345
15	University_holidays 6.951928 0.047321 0.008236
16	low_temp 4.153927 0.028325 0.004914
17	high_temp 3.914921 0.026610 0.004651
18	high_wind 3.279167 0.022286 0.003879
19	abnormal_rain 3.197763 0.021756 0.003783

Python 3

See the whole table with `table.as_data_frame()`  
Now tuning drf Grid Build progress: [██████████] 100%

0 .....

1 .....

2 .....

3 .....

4 .....

5 .....

6 .....

7 .....

8 .....

9 .....

10 .....

11 .....

12 .....

13 .....

14 .....

15 .....

16 .....

17 .....

18 .....

19 .....

Mode: Command Ln 1, Col 1 Mail\_football\_prediction\_skeleton.ipynb

ENG 6:33 PM 11/10/2020

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

Python 3

18 ... 0.00820192451792715  
19 ... 0.008473527510839192

[20 rows x 7 columns]

Model2MetricRegression.drf  
\*\* Reported on test data. \*\*

MSE: 0.00620581480906411  
RMSE: 0.078779941611988  
MAE: 0.052892439326487314  
MSEL: 0.0543676767680003704  
Mean Residual Deviance: 0.00620581480906411

drf prediction progress: [██████████] 100%  
drf Grid Build progress: [██████████] 100%  
stackeddense Model Build progress: [██████████] 100%  
Best Base-learner Test MSE: 0.00620581480906411  
0.00620581480906411  
stackeddense prediction progress: [██████████] 100%  
stackeddense Model Build progress: [██████████] 100%  
Best Base-learner Test MSE: 0.0059873316895677155  
Ensemble Test MSE: 0.00611763224183206  
stackeddense prediction progress: [██████████] 100%

Python 3 | idle

Mode: Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb

EN 6:33 PM INTL 11/10/2020

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Python 3

```
[54]: h2o_model_1 = h2o.model(data.outliers_removed.dataset_type='OutlierRemoved')
loss_h2o_model_outlier_removed = h2o_model_1.train_and_evaluate_models()
loss_values_all_models.append(loss_h2o_model_outlier_removed)

Defined 2 models
Parse progress: [██████████] 100%
Parse progress: [██████████] 100%
gbm Model Build progress: [██████████] 100%

ModelMetricsRegression: gbm
** Reported on test data. **

MSE: 0.00077860712499883
RMSE: 0.072339582362956
MAE: 0.0533722874811195
RMSLE: 0.0544012542785118
Mean Residual Deviance: 0.0002789697112499883

gbm prediction progress: [██████████] 100%
=====
H2OGradientBoosting
default H2O model fitting...
Model Details
=====
H2OGradientBoostingEstimator : Gradient Boosting Machine
Model Key: GBM_model_python_1664675476624_3

Model Summary:
  number_of_trees number_of_internal_trees model_size_in_bytes min_depth max_depth mean_depth min_leaves max_leaves mean_leaves
0          100.0                 100.0            36709.0      5.0       5.0       5.0     32.0      10.0      24.53

ModelMetricsRegression: gbm
** Reported on train data. **

MSE: 0.0010165291662191643
RMSE: 0.031883054538088151
MAE: 0.022800957239908
RMSLE: 0.023485028454244097
```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 | idle

Windows Taskbar: Google, Chrome, Microsoft Edge, PowerShell, Firefox, Excel, Word, Google, BITS Pilani - Login, ACT Fibernet Portal..., Customer Experience..., AIML\_SK Links, Django: Web frame..., AI - Updates, BITS-Wheebox, BerkeleyX-AI OL Co... 6:34 PM INTL 11/10/2020

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Python 3

```
RMSLE: 0.023485028454244967
Mean Residual Deviance: 0.00010165291662191643

ModelMetricsRegression: gbm
** Reported on validation data. **

MSE: 0.000530641092069243
RMSE: 0.07593759785646955
MAE: 0.05023597905840224
RMSLE: 0.053375462085237954
Mean Residual Deviance: 0.000530641092069243

ModelMetricsRegression: gbm
** Reported on cross-validation data. **

MSE: 0.00518223979558678
RMSE: 0.07147876553472133
MAE: 0.04905536460455974
RMSLE: 0.05192761608842768
Mean Residual Deviance: 0.00518223979558678

Cross-Validation Metrics Summary:
  mean_sd_cv_1_valid_cv_2_valid_cv_3_valid_cv_4_valid_cv_5_valid
0  mae 0.04906113 0.0023015528 0.0595769 0.0480005 0.04831437 0.0481627 0.05191021
1  mean_residual_deviance 0.0051097935 5.822343E-4 0.057576925 0.0049935158 0.0044443 0.0049906 0.0056236997
2  mse 0.0051097935 5.822343E-4 0.057576925 0.0049935158 0.0044443 0.0049906 0.0056236997
3  r2  0.41249484 0.052671965 0.47723487 0.35058867 0.39135644 0.38688716 0.4566091
4  residual_deviance 0.0051097935 5.822343E-4 0.057576925 0.0049935158 0.0044443 0.0049906 0.0056236997
5  rmse 0.071390098 0.0040695574 0.0757979595 0.068390238 0.06666558 0.07084418 0.07532304
6  rmsle 0.05185107 0.0031815784 0.055355046 0.0490919 0.048135735 0.05090974 0.05490204

Scoring History:
  timestamp duration number_of_trees training_rmse training_mae training_deviance validation_rmse validation_mae validation_deviance
0  2020-11-05 16:16:48 7.405 sec 0.0 0.093965 0.067588 0.008830 0.098720 0.069696 0.009746
1  2020-11-06 16:16:48 7.422 sec 1.0 0.089791 0.064603 0.008062 0.095479 0.066333 0.009116
2  2020-11-04 16:16:48 7.414 sec 2.0 0.084150 0.061658 0.007422 0.093361 0.064074 0.008531
```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 | idle

Windows Taskbar: Google, Chrome, Microsoft Edge, PowerShell, Firefox, Excel, Word, Google, BITS Pilani - Login, ACT Fibernet Portal..., Customer Experience..., AIML\_SK Links, Django: Web frame..., AI - Updates, BITS-Wheebox, BerkeleyX-AI OL Co... 6:35 PM INTL 11/10/2020

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Launcher Mail\_footfall\_prediction\_ske

Scoring History:

	timestamp	duration	number_of_trees	training_rmse	training_mae	training_deviance	validation_rmse	validation_mae	validation_deviance
0	2020-11-06 16:16:48	7.405 sec	0.0	0.093965	0.067588	0.008830	0.098720	0.068686	0.009746
1	2020-11-06 16:16:48	7.422 sec	1.0	0.089791	0.064043	0.008002	0.095479	0.068333	0.009116
2	2020-11-06 16:16:48	7.434 sec	2.0	0.086150	0.061958	0.007422	0.092381	0.064074	0.008331
3	2020-11-06 16:16:48	7.448 sec	3.0	0.082961	0.059700	0.006883	0.090065	0.062080	0.008112
4	2020-11-06 16:16:48	7.454 sec	4.0	0.080203	0.057637	0.006432	0.087943	0.060582	0.007734
5	2020-11-06 16:16:48	7.479 sec	5.0	0.077829	0.055960	0.006057	0.085783	0.059157	0.007359
6	2020-11-06 16:16:48	7.498 sec	6.0	0.075847	0.054330	0.005722	0.083823	0.057801	0.007026
7	2020-11-06 16:16:48	7.511 sec	7.0	0.073714	0.052871	0.005434	0.082148	0.056562	0.006748
8	2020-11-06 16:16:48	7.526 sec	8.0	0.071738	0.051473	0.005146	0.081148	0.055370	0.006585
9	2020-11-06 16:16:48	7.540 sec	9.0	0.070139	0.050371	0.004920	0.080661	0.055151	0.006506
10	2020-11-06 16:16:48	7.555 sec	10.0	0.068427	0.049085	0.004682	0.079614	0.054540	0.006370
11	2020-11-06 16:16:48	7.571 sec	11.0	0.067076	0.047995	0.004499	0.078930	0.053744	0.006230
12	2020-11-06 16:16:48	7.588 sec	12.0	0.065750	0.047039	0.004323	0.078391	0.053170	0.006145
13	2020-11-06 16:16:48	7.598 sec	13.0	0.065584	0.046167	0.004171	0.077441	0.052437	0.006028
14	2020-11-06 16:16:48	7.607 sec	14.0	0.065216	0.045021	0.003996	0.076852	0.051838	0.005906
15	2020-11-06 16:16:48	7.616 sec	15.0	0.061801	0.044231	0.003819	0.076789	0.051863	0.005997
16	2020-11-06 16:16:48	7.628 sec	16.0	0.060975	0.043465	0.003718	0.076260	0.051384	0.005816
17	2020-11-06 16:16:48	7.637 sec	17.0	0.060166	0.042869	0.003620	0.076137	0.051146	0.005797
18	2020-11-06 16:16:48	7.645 sec	18.0	0.059309	0.042269	0.003518	0.075683	0.050861	0.005728
19	2020-11-06 16:16:48	7.659 sec	19.0	0.058505	0.041543	0.003423	0.075234	0.050523	0.005660

See the whole table with `table.as_data_frame()`

Variable Importances:

variable	relative_importance	scaled_importance	percentage
0	Weekday	13.792759	1.000000 0.238900

Python 3 | idle Mode: Command L1, Col 1 Mail\_footfall\_prediction\_skeleton.ipynb ENG INTEL 11/10/2020

Google JupyterLab localhost:8888/lab

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Launcher Mail\_footfall\_prediction\_ske Python 3

18 2020-11-06 16:16:48 7.845 sec 18.0 0.059309 0.042269 0.003518 0.075683 0.050861 0.005728  
19 2020-11-06 16:16:48 7.659 sec 19.0 0.058505 0.041548 0.003423 0.075234 0.050526 0.005660

See the whole table with `table.as_data_frame()`

Variable Importances:

	variable	relative_importance	scaled_importance	percentage
0	Weekday	13.782759	1.000000	0.23900
1	InCountTotal_lag1	6.770346	0.491219	0.117352
2	Day	6.344209	0.469300	0.109965
3	Weekend	5.567885	0.403975	0.096509
4	InCountTotal_lag3	4.368818	0.316614	0.075639
5	InCountTotal_lag5	3.613246	0.262157	0.062629
6	InCountTotal_lag7	3.419177	0.24076	0.059265
7	InCountTotal_lag6	2.705990	0.196324	0.046902
8	InCountTotal_lag2	2.654212	0.192575	0.046006
9	mean_temp	2.537713	0.184122	0.043987
10	InCountTotal_lag4	1.864299	0.135263	0.032314
11	rain	1.037118	0.075248	0.017977
12	wind_speed	0.756545	0.054891	0.013113
13	Month	0.754096	0.054713	0.013071
14	School_holidays	0.642702	0.046531	0.011140
15	University_holidays	0.512526	0.037180	0.008884
16	UKBankHoliday	0.321535	0.023329	0.005573
17	high_wind	0.028969	0.002102	0.000502
18	abnormal_rain	0.009033	0.000655	0.000157
19	low_temp	0.005442	0.000395	0.000094

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```

See the whole table with table.as_data_frame()
Now tuning gbm Grid Build progress: [██████████] 100%
gbm col_sample_rate ... model_ids mse
0 0.30000000000000004 ... H2GradientBoosting_model_18 0.005078836010218611
1 0.4 ... H2GradientBoosting_model_8 0.0051088579111282
2 0.5 ... H2GradientBoosting_model_17 0.0051229077450409
3 0.9 ... H2GradientBoosting_model_1 0.00517597515947249
4 0.8 ... H2GradientBoosting_model_20 0.00518653787554791
5 0.5 ... H2GradientBoosting_model_13 0.005192807407250871
6 0.9 ... H2GradientBoosting_model_16 0.0051982573952441
7 0.4 ... H2GradientBoosting_model_10 0.0052001241114489
8 0.5000000000000001 ... H2GradientBoosting_model_2 0.005209184608577699
9 0.1 ... H2GradientBoosting_model_14 0.005254315765298267
10 0.2 ... H2GradientBoosting_model_5 0.005385572187079125
11 0.9 ... H2GradientBoosting_model_19 0.005433537355577598
12 0.2 ... H2GradientBoosting_model_9 0.00545433537355577598
13 0.7000000000000001 ... H2GradientBoosting_model_12 0.005751981136713425
14 0.5 ... H2GradientBoosting_model_4 0.005779812426848832
15 0.5 ... H2GradientBoosting_model_1 0.006122159312091363
16 0.4 ... H2GradientBoosting_model_17 0.0061422159312091363
17 0.5 ... H2GradientBoosting_model_6 0.00627236941199247
18 0.5 ... H2GradientBoosting_model_15 0.006581127869412801
19 1.0 ... H2GradientBoosting_model_3 0.007038581441410625

[28 rows x 7 columns]

ModelMetricsRegression: gbm
** Reported on test data. **

MSE: 0.0063345365250468575
RMSE: 0.0795898016397084
MAE: 0.052120038929593424
MSLE: 0.0522915588474274
RMSLE: 0.056251121123378
Mean Residual Deviance: 0.0063345365250468575

gbm prediction progress: [██████████] 100%
gbm Model Build progress: [██████████] 100%
stackedensemble Model Build progress: [██████████] 100%
Best Base-learner Test MSE: 0.006120031090418484
Ensemble Test MSE: 0.006183681975318625
stackedensemble prediction progress: [██████████] 100%

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3 idle

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```

enmemize('test', 'test', 'enmemized')
stackedensemble prediction progress: [██████████] 100%
drf Model Build progress: [██████████] 100%

ModelMetricsRegression: drf
** Reported on test data. **

MSE: 0.0062710794404340374
RMSE: 0.0721984742652842
MAE: 0.0522915588474274
MSLE: 0.0564112423707231
Mean Residual Deviance: 0.0062710794404340374

drf prediction progress: [██████████] 100%
*****
H2RandomForest
default H2o model fitting...
Model Key: DRF
*****
H2RandomForestEstimator : Distributed Random Forest
Model Key: DRF_model_python_1604675476624_4

Model Summary:
  number_of_trees  number_of_internal_trees  model_size_in_bytes  min_depth  max_depth  mean_depth  min_leaves  max_leaves  mean_leaves
0           100.0                  100.0            1002184.0      20.0       200     200       644.0      916.0      793.27

ModelMetricsRegression: drf
** Reported on train data. **

MSE: 0.0052545087768113966
RMSE: 0.07248799856405114
MAE: 0.052297777227575
MSLE: 0.0514091818913844
Mean Residual Deviance: 0.0052545087768113966

ModelMetricsRegression: drf
** Reported on validation data. **

MSE: 0.00515895288729364
RMSE: 0.0718045625899174
MAE: 0.04744646951917641

```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3 idle

Google JupyterLab localhost:8888/lab

File Edit View Run Kernel Tabs Settings Help

Launcher Mail\_footfall\_prediction\_ske

```

MAE: 0.0474446895191764
RMSE: 0.05132977133688185
Mean Residual Deviance: 0.005155895288729364

ModellingMetricsRegression: drf
** Reported on cross-validation data. **

MSE: 0.0052686820400688475
RMSE: 0.02725827375750537
MAE: 0.04884369552777601
RMSE: 0.052452106886560336
Mean Residual Deviance: 0.0052068820400688475

Cross-Validation Metrics Summary:
  mean sd cv_1_valid cv_2_valid cv_3_valid cv_4_valid cv_5_valid
0 mae 0.04805333 0.0036033657 0.052259443 0.045606233 0.045281082 0.04819288 0.05392481
1 mean_residual_deviance 0.005206884774 0.0010062407 0.0062489645 0.0041930878 0.00429978 0.005049812 0.006230545
2 mse 0.005206884774 0.0010062407 0.0062489645 0.0041930878 0.00429978 0.005049812 0.006230545
3 r2 0.40844602 0.020102 0.4326592 0.4196492 0.4111235 0.3796127 0.4021765
4 residual_deviance 0.005206884774 0.0010062407 0.0062489645 0.0041930878 0.00429978 0.005049812 0.006230545
5 rmse 0.07190022 0.006957452 0.0790539 0.06754054 0.06574214 0.07106203 0.07906039
6 rmse 0.05257777 0.005185374 0.057349088 0.0474031 0.04737675 0.051180488 0.05808851

Scoring History:
  timestamp duration number_of_trees training_rmse training_mae training_deviance validation_rmse validation_mae validation_deviance
0 2020-11-06 16:18:50 15.09 sec 0.0 NaN NaN NaN NaN NaN
1 2020-11-06 16:18:50 15.139 sec 1.0 0.108703 0.075464 0.011385 0.101125 0.068554 0.010226
2 2020-11-06 16:18:50 15.180 sec 2.0 0.102648 0.071132 0.010537 0.089144 0.060568 0.007947
3 2020-11-06 16:18:51 15.243 sec 3.0 0.096694 0.067673 0.009740 0.079948 0.053145 0.006392
4 2020-11-06 16:18:51 15.249 sec 4.0 0.094420 0.065627 0.008915 0.077376 0.052694 0.005987
5 2020-11-06 16:18:51 15.293 sec 5.0 0.091838 0.064073 0.008434 0.076545 0.052281 0.005859
6 2020-11-06 16:18:51 15.317 sec 6.0 0.090564 0.063275 0.008202 0.075589 0.052103 0.005714
```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 idle

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Launcher Mail\_footfall\_prediction\_ske

```

5 2020-11-06 16:18:51 15.293 sec 5.0 0.091838 0.064073 0.008434 0.076545 0.052281 0.005859
6 2020-11-06 16:18:51 15.317 sec 6.0 0.090564 0.063275 0.008202 0.075589 0.052103 0.005714
7 2020-11-06 16:18:51 15.341 sec 7.0 0.088319 0.061935 0.007800 0.075005 0.051199 0.005626
8 2020-11-06 16:18:51 15.365 sec 8.0 0.084732 0.059433 0.007179 0.074561 0.050725 0.005559
9 2020-11-06 16:18:51 15.389 sec 9.0 0.083521 0.058476 0.006978 0.074016 0.050299 0.005478
10 2020-11-06 16:18:51 15.413 sec 10.0 0.082669 0.057843 0.006837 0.074677 0.050778 0.005577
11 2020-11-06 16:18:51 15.439 sec 11.0 0.0811831 0.057055 0.006696 0.074230 0.050048 0.005510
12 2020-11-06 16:18:51 15.462 sec 12.0 0.080049 0.056266 0.006408 0.073910 0.049878 0.005463
13 2020-11-06 16:18:51 15.488 sec 13.0 0.079996 0.056161 0.006399 0.073522 0.049715 0.005405
14 2020-11-06 16:18:51 15.512 sec 14.0 0.079815 0.056170 0.006370 0.073612 0.049559 0.005419
15 2020-11-06 16:18:51 15.552 sec 15.0 0.079789 0.056130 0.006366 0.073298 0.049332 0.005373
16 2020-11-06 16:18:51 15.578 sec 16.0 0.079566 0.055923 0.006331 0.073841 0.049469 0.005453
17 2020-11-06 16:18:51 15.602 sec 17.0 0.078893 0.055389 0.006224 0.073795 0.049448 0.005446
18 2020-11-06 16:18:51 15.626 sec 18.0 0.078897 0.055305 0.006235 0.073910 0.049355 0.005443
19 2020-11-06 16:18:51 15.652 sec 19.0 0.078338 0.054707 0.006137 0.073279 0.049700 0.005370

See the whole table with table.as_data_frame()

Variable Importances:
  variable relative_importance scaled_importance percentage
0 Weekday 136.721832 1.000000 0.170930
1 InCountTotal_Jag1 83.173897 0.068344 0.103984
2 InCountTotal_Jag7 79.707718 0.582992 0.099651
3 WeekIn 72.392471 0.529487 0.090505
4 Day 54.921387 0.401702 0.088663
5 InCountTotal_Jag3 52.086010 0.380963 0.065118
6 InCountTotal_Jag5 51.890717 0.379535 0.064874
7 InCountTotal_Jag7 51.560174 0.377177 0.064474
```

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.pyrb

Python 3 idle

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7 InCountTotal\_lag5 51.568356 0.377177 0.064471  
8 InCountTotal\_lag2 40.5977359 0.296934 0.050755  
9 mean\_temp 39.309521 0.287515 0.049145  
10 InCountTotal\_lag4 31.473110 0.230198 0.039348  
11 Month 27.853939 0.203727 0.034623  
12 wind\_speed 26.291880 0.192302 0.032870  
13 rain 18.123748 0.132374 0.022661  
14 School\_holidays 9.674769 0.070762 0.012095  
15 University\_holidays 7.253795 0.053055 0.009069  
16 low\_temp 4.050148 0.029623 0.005064  
17 high\_temp 3.934752 0.028779 0.004919  
18 UKBankHoliday 3.405069 0.024872 0.004251  
19 high\_wind 2.810534 0.020557 0.003514

See the whole table with `table.as_data_frame()`

Now tuning

drf Grid Build progress: [██████████] 100%

0 ... 0.0052525484850568165  
1 ... 0.00560380989039224656  
2 ... 0.0051609915880159819  
3 ... 0.005164519210376568869  
4 ... 0.006253839340525964  
5 ... 0.006297597748026385  
6 ... 0.00630000000000000  
7 ... 0.006437791245552392  
8 ... 0.006501111594343689  
9 ... 0.006515833826568093  
10 ... 0.00708980159646373  
11 ... 0.007101359532056231  
12 ... 0.007206103761550721  
13 ... 0.007515952756250934  
14 ... 0.007525322342502378  
15 ... 0.007525322342502378  
16 ... 0.007525322342502378  
17 ... 0.0075248221134094  
18 ... 0.0075248221134094  
19 ... 0.008011983040151161

[28 rows x 7 columns]

ModelMetricsRegression: drf  
\*\* Reported on test data. \*\*

MSE: 0.0061698898651121  
RMSE: 0.078545355456432974  
MAE: 0.0515966452878627  
MSLE: 0.05601285179000048  
Mean Residual Deviance: 0.006169889806511321

drf prediction progress: [██████████] 100%

drf Grid Build progress: [██████████] 100%

stackedensemble Model Build progress: [██████████] 100%

Best Base-learner Test MSE: 0.0063508980806511321  
Ensemble Test MSE: 0.006182324998759835

stackedensemble prediction progress: [██████████] 100%

stackedensemble Model Build progress: [██████████] 100%

Best Base-learner Test MSE: 0.0062710794494348374  
Ensemble Test MSE: 0.0061464889363274614

stackedensemble prediction progress: [██████████] 100%

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3 | idle

Windows Taskbar: Google, Chrome, Microsoft Edge, Firefox, Excel, Word, Google, BITS Pilani - Login, ACT Fibernet Portal..., Customer Experience..., AIML\_SK Links, Django: Web frame..., AI- Updates, BITS-Wheebox, BerkeleyX-AI OL Co...  
ENG 6:40 PM INTL 11/10/2020

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Launcher Mail\_footfall\_prediction\_ske

Python 3

16 ... 0.007525322342502378  
17 ... 0.0075248221134094  
18 ... 0.0075248221134094  
19 ... 0.008011983040151161

[28 rows x 7 columns]

ModelMetricsRegression: drf  
\*\* Reported on test data. \*\*

MSE: 0.0061698898651121  
RMSE: 0.078545355456432974  
MAE: 0.0515966452878627  
MSLE: 0.05601285179000048  
Mean Residual Deviance: 0.006169889806511321

drf prediction progress: [██████████] 100%

drf Grid Build progress: [██████████] 100%

stackedensemble Model Build progress: [██████████] 100%

Best Base-learner Test MSE: 0.0063508980806511321  
Ensemble Test MSE: 0.006182324998759835

stackedensemble prediction progress: [██████████] 100%

stackedensemble Model Build progress: [██████████] 100%

Best Base-learner Test MSE: 0.0062710794494348374  
Ensemble Test MSE: 0.0061464889363274614

stackedensemble prediction progress: [██████████] 100%

Mode Command Ln 1, Col 1 Mail\_footfall\_prediction\_skeleton.py

Python 3 | idle

Windows Taskbar: Google, Chrome, Microsoft Edge, Firefox, Excel, Word, Google, BITS Pilani - Login, ACT Fibernet Portal..., Customer Experience..., AIML\_SK Links, Django: Web frame..., AI- Updates, BITS-Wheebox, BerkeleyX-AI OL Co...  
ENG 6:40 PM INTL 11/10/2020

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Project / Capstone-Project-master /

- Name
- data
- AIML Capstone - Gro...
- Coverage.xlsx
- FullDBSCAN.ipynb
- Mail\_footfall\_predict...**
- Mail\_footfall\_predict...
- preprocess.py
- Project Plan.xlsx
- README.md
- visualization.py
- work\_flow.pptx

Launcher Code Python 3

```
[55]: loss_h2o_model_outlier_removed.sort_values('mape').head(25)
```

	Model_type	mape	mse	me	mae	mpre	rmse	R2 score	Adj R2 score	AIC	BIC
4	H2ORandomForest_OutlierRemoved_rscv	22.91991	0.06169	-0.19140	0.051510	0.078740	0.078544	0.337415	0.294793	-1813.94456	-1727.96717
5	H2ORandomForest_OutlierRemoved_random_grid_ensemble	23.16026	0.06182	-0.19210	0.051885	0.078467	0.078628	0.335991	0.293277	-1812.410746	-1726.613004
6	H2O_Stacked_Ensemble_OutlierRemoved_ensemble	23.170633	0.06145	-0.19348	0.051925	0.078054	0.078400	0.335943	0.297377	-1814.534087	-1728.736345
0	H2OGradientBoosting_OutlierRemoved_random_grid_ensemble	23.45252	0.06184	-0.16530	0.052624	0.079992	0.078568	0.335848	0.282322	-1806.750525	-1720.952784
2	H2OGradientBoosting_OutlierRemoved_rscv	23.482069	0.06035	-0.17766	0.052128	0.084799	0.079590	0.319646	0.275880	-1803.534804	-1717.736863
1	H2OGradientBoosting_OutlierRemoved_rscv	23.601468	0.060271	-0.18408	0.052292	0.085424	0.079190	0.326461	0.283134	-1807.209479	-1721.411737
3	H2ORandomForest_OutlierRemoved_default_mode	23.601468	0.060271	-0.18408	0.052292	0.085424	0.079190	0.326461	0.283134	-1807.209479	-1721.411737

7.4 Model training with AutoML

```
[56]: auto_ml_model = h2o_model(data_final,dataset_type='Normal')
loss_auto_ml_model = auto_ml_model.autoML(max_models=30,max_runtime_secs=1000)
loss_values_all_models.append(loss_auto_ml_model)

Parse progress: [██████████] 100%
Parse progress: [██████████] 100%
AutoML progress: [██████████] 100%
16:28:54.71: User specified a validation frame with cross-validation still enabled. Please note that the models will still be validated using cross-validation only, the validation frame will be used to provide purely informative validation metrics on the trained models.
```

model_id	mean_residual_deviance	rmse	mse	mae	rmsele
StackedEnsemble_BestOfFamily_AutoML_20201106_162054	0.00604146	0.0777268	0.06061446	0.0512674	0.0553327
XRT_1_AutoML_20201106_162054	0.00605452	0.0778108	0.0605452	0.0518174	0.05597723
GBM_1_AutoML_20201106_162054	0.00614575	0.0783948	0.00514575	0.0518768	0.0576568
StackedEnsemble_AllModels_AutoML_20201106_162054	0.00617715	0.0785948	0.00617715	0.0520592	0.06020273
GBM_3_AutoML_20201106_162054	0.006196	0.0787147	0.006196	0.0519991	0.056008
GBM_2_AutoML_20201106_162054	0.00619931	0.0787357	0.00619931	0.0525321	0.056339
GBT_1_AutoML_20201106_162054	0.00643203	0.0788704	0.00643203	0.0518473	0.0545107

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- Name
- data
- AIML Capstone - Gro...
- Coverage.xlsx
- FullDBSCAN.ipynb
- Mail\_footfall\_predict...**
- Mail\_footfall\_predict...
- preprocess.py
- Project Plan.xlsx
- README.md
- visualization.py
- work\_flow.pptx

Launcher Code Python 3

```
[56]: GBM_3_AutoML_20201106_162054
GBM_2_AutoML_20201106_162054
GBM_1_AutoML_20201106_162054
DRF_1_AutoML_20201106_162054
XGBoost_grid_1_AutoML_20201106_162054_modeL6
GBM_grid_1_AutoML_20201106_162054_modeL4
GBM_grid_1_AutoML_20201106_162054
GBM_grid_1_AutoML_20201106_162054_modeL5
GBM_grid_1_AutoML_20201106_162054_modeL3
GBM_grid_1_AutoML_20201106_162054
GBM_5_AutoML_20201106_162054
GBM_grid_1_AutoML_20201106_162054_modeL2
XGBoost_grid_1_AutoML_20201106_162054_modeL4
XGBoost_1_AutoML_20201106_162054
XGBoost_grid_1_AutoML_20201106_162054_modeL2
XGBoost_grid_1_AutoML_20201106_162054_modeL1
XGBoost_grid_1_AutoML_20201106_162054_modeL7
GBM_grid_1_AutoML_20201106_162054_modeL1
XGBoost_grid_1_AutoML_20201106_162054_modeL9
XGBoost_grid_1_AutoML_20201106_162054_modeL8
XGBoost_2_AutoML_20201106_162054
XGBoost_grid_1_AutoML_20201106_162054_modeL5
DeepLearning_grid_2_AutoML_20201106_162054_modeL1
DeepLearning_grid_3_AutoML_20201106_162054_modeL1
XGBoost_1_AutoML_20201106_162054
DeepLearning_grid_1_AutoML_20201106_162054_modeL2
DeepLearning_1_AutoML_20201106_162054
0.006196 0.0787147 0.006196 0.0519991 0.056008
0.00619931 0.0787357 0.00619931 0.0525321 0.056339
0.00623000 0.0789306 0.00623000 0.0518157 0.0562195
0.00632800 0.0795538 0.00632800 0.0536711 0.05669931
0.00649848 0.0806132 0.00649848 0.0530953 0.0575134
0.00650937 0.0806097 0.00650937 0.0542919 0.057712
0.00651985 0.0813813 0.00651985 0.0537344 0.0580231
0.0065101 0.0815537 0.0065101 0.0545145 0.0581541
0.00666608 0.0816498 0.00666608 0.0535518 0.0583741
0.0067132 0.0819341 0.0067132 0.0561555 0.0584218
0.00673721 0.0820805 0.00673721 0.0568375 0.0585725
0.00674128 0.0821053 0.00674128 0.0569568 0.0586813
0.00687686 0.0829255 0.00687686 0.0579128 0.0592073
0.00700648 0.0837056 0.00700648 0.0576526 0.0599003
0.00706011 0.0840245 0.00706011 0.0571254 0.0602020
0.00707059 0.0840556 0.00707059 0.0576915 0.0598428
0.00719808 0.0848458 0.00719808 0.0577816 0.0600545
0.00739456 0.0859916 0.00739456 0.0607357 0.0618995
0.0074959 0.0865789 0.0074959 0.0601863 0.0620207
0.00751451 0.0866844 0.00751451 0.0589449 0.0620731
0.00754994 0.0868904 0.00754994 0.0589508 0.0624095
0.00774469 0.0880153 0.00774469 0.0598464 0.062712
0.0081097 0.0895024 0.0081097 0.0597412 0.0641131
0.00806302 0.0897787 0.00806302 0.0583280 0.0646126
0.00816128 0.0903398 0.00816128 0.0635224 0.0645316
0.00823815 0.0907948 0.00823825 0.0657851 0.0651732
0.00824030 0.0913746 0.00824030 0.0660604 0.0652321
```

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DeepLearning\_grid\_1\_AutoML\_20201106\_162054\_modelL2 0.00816128 0.0903998 0.00816128 0.0635224 0.0648316  
DeepLearning\_LAutoML\_20201106\_162054 0.00823825 0.0907848 0.00823825 0.0657961 0.0651732  
GLM\_1\_AutoML\_20201106\_162054 0.008690638 0.0945736 0.008690638 0.0686886 0.0673304  
DeepLearning\_grid\_1\_AutoML\_20201106\_162054\_modelL1 0.0089483 0.0945955 0.0089483 0.0649044 0.0681545

Model Details  
\*\*\*\*\*  
StackedEnsembleEstimator : Stacked Ensemble  
Model Key: StackedEnsemble\_BestOfFamily\_AutoML\_20201106\_162054

Model summary for this model

ModelMetricsRegression(M): stackedensemble  
\*\* Reported on train data. \*\*

MSE: 0.0018078327470448587  
RMSE: 0.0321575883203847255  
MAE: 0.023799832358467  
MSLE: 0.023799884554463  
R2: 0.88547491101313  
P2: 0.88547491101313  
AIC: 0.8018478327470448587  
Null degrees of freedom: 1413  
Residual degrees of freedom: 1408  
Null Deviance: 13.04956292466353  
Residual Deviance: 1.4805043843214303  
AIC: -567.820857866416

ModelMetricsRegression(M): stackedensemble  
\*\* Reported on validation data. \*\*

MSE: 0.00487490169040242595  
RMSE: 0.0638340566401923  
MAE: 0.043788355173195135  
MSLE: 0.043788355173195135  
R2: 0.55157198776417972  
Mean Residual Deviance: 0.0040749816984842595  
Null degrees of freedom: 347  
Residual degrees of freedom: 342  
Null Deviance: 2.93809991378214  
Residual Deviance: 1.4198865782686822  
AIC: -913.4389937465318

