

# Train a model with bike rental data using XGBoost algorithm

## Training log1p(count) dataset

Model is trained with XGBoost installed in notebook instance

In the later examples, we will train using SageMaker's XGBoost algorithm

```
In [1]: # Install xgboost in notebook instance.  
##### Command to install xgboost  
!pip install xgboost
```

```
Looking in indexes: https://pypi.org/simple, https://pip.repos.neuron.amazonaws.com  
Requirement already satisfied: xgboost in /home/ec2-user/anaconda3/envs/python3/lib/python3.10/site-packages (1.7.6)  
Requirement already satisfied: numpy in /home/ec2-user/anaconda3/envs/python3/lib/python3.10/site-packages (from xgb  
oost) (1.22.3)  
Requirement already satisfied: scipy in /home/ec2-user/anaconda3/envs/python3/lib/python3.10/site-packages (from xgb  
oost) (1.10.1)
```

```
In [2]: import sys  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
from sklearn.metrics import mean_squared_error, mean_absolute_error  
  
# XGBoost  
import xgboost as xgb
```

```
In [3]: column_list_file = 'bike_train_column_list.txt'  
train_file = 'bike_train.csv'  
validation_file = 'bike_validation.csv'  
test_file = 'bike_test.csv'
```

```
In [4]: columns = ''  
with open(column_list_file, 'r') as f:  
    columns = f.read().split(',')
```

In [5]: columns

Out[5]: ['count',  
'season',  
'holiday',  
'workingday',  
'weather',  
'temp',  
'atemp',  
'humidity',  
'windspeed',  
'year',  
'month',  
'day',  
'dayofweek',  
'hour']

In [6]: *# Specify the column names as the file does not have column header*  
df\_train = pd.read\_csv(train\_file,names=columns)  
df\_validation = pd.read\_csv(validation\_file,names=columns)

In [7]: df\_train.head()

Out[7]:

	count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
0	4.477337	3	0	0	2	26.24	30.305	73	7.0015	2011	9	3	5	0
1	5.517453	3	0	1	1	32.80	34.850	33	7.0015	2012	8	13	0	14
2	5.814131	4	0	0	1	15.58	19.695	40	11.0014	2011	11	5	5	17
3	6.436150	3	0	1	1	32.80	37.880	55	12.9980	2012	8	9	3	19
4	4.262680	2	0	1	1	13.94	17.425	76	7.0015	2011	4	14	3	6

In [8]: df\_validation.head()

```
Out[8]:
```

	count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
0	6.095825	3	0	1	2	28.70	33.335	79	12.9980	2011	7	7	3	8
1	5.961005	2	0	0	1	32.80	37.880	55	12.9980	2011	6	11	5	13
2	1.098612	1	0	1	1	14.76	16.665	40	19.9995	2011	2	14	0	2
3	3.891820	1	0	1	1	9.02	9.090	47	36.9974	2011	2	8	1	10
4	4.025352	4	0	0	1	10.66	15.150	87	0.0000	2011	12	4	6	8

```
In [9]: X_train = df_train.iloc[:,1:] # Features: 1st column onwards
y_train = df_train.iloc[:,0].ravel() # Target: 0th column

X_validation = df_validation.iloc[:,1:]
y_validation = df_validation.iloc[:,0].ravel()
```

```
In [10]: # XGBoost Training Parameter Reference:
# https://github.com/dmlc/xgboost/blob/master/doc/parameter.md
#regressor = xgb.XGBRegressor(max_depth=5,eta=0.1,subsample=0.7,num_round=150)
regressor = xgb.XGBRegressor(max_depth=5,n_estimators=150)
```

```
In [11]: regressor
```

```
Out[11]:
```

▼ XGBRegressor

XGBRegressor(base\_score=None, booster=None, callbacks=None,  
           colsample\_bylevel=None, colsample\_bynode=None,  
           colsample\_bytree=None, early\_stopping\_rounds=None,  
           enable\_categorical=False, eval\_metric=None, feature\_types=None,  
           gamma=None, gpu\_id=None, grow\_policy=None, importance\_type=None,  
           interaction\_constraints=None, learning\_rate=None, max\_bin=None,  
           max\_cat\_threshold=None, max\_cat\_to\_onehot=None,  
           max\_delta\_step=None, max\_depth=5, max\_leaves=None,  
           min\_child\_weight=None, missing=nan, monotone\_constraints=None,

```
In [12]: regressor.fit(X_train,y_train, eval_set = [(X_train, y_train), (X_validation, y_validation)])
```

[0]	validation_0-rmse:3.06460	validation_1-rmse:3.07128
[1]	validation_0-rmse:2.18561	validation_1-rmse:2.18977
[2]	validation_0-rmse:1.57973	validation_1-rmse:1.58729
[3]	validation_0-rmse:1.15916	validation_1-rmse:1.16584
[4]	validation_0-rmse:0.86291	validation_1-rmse:0.87348
[5]	validation_0-rmse:0.67593	validation_1-rmse:0.68940
[6]	validation_0-rmse:0.55710	validation_1-rmse:0.57295
[7]	validation_0-rmse:0.47210	validation_1-rmse:0.49133
[8]	validation_0-rmse:0.41153	validation_1-rmse:0.43610
[9]	validation_0-rmse:0.37590	validation_1-rmse:0.40111
[10]	validation_0-rmse:0.34717	validation_1-rmse:0.37443
[11]	validation_0-rmse:0.33554	validation_1-rmse:0.36382
[12]	validation_0-rmse:0.32311	validation_1-rmse:0.35253
[13]	validation_0-rmse:0.31561	validation_1-rmse:0.34656
[14]	validation_0-rmse:0.31065	validation_1-rmse:0.34307
[15]	validation_0-rmse:0.30379	validation_1-rmse:0.33729
[16]	validation_0-rmse:0.30120	validation_1-rmse:0.33574
[17]	validation_0-rmse:0.29738	validation_1-rmse:0.33265
[18]	validation_0-rmse:0.29363	validation_1-rmse:0.33012
[19]	validation_0-rmse:0.28230	validation_1-rmse:0.31960
[20]	validation_0-rmse:0.28065	validation_1-rmse:0.31865
[21]	validation_0-rmse:0.27929	validation_1-rmse:0.31837
[22]	validation_0-rmse:0.27543	validation_1-rmse:0.31537
[23]	validation_0-rmse:0.27240	validation_1-rmse:0.31359
[24]	validation_0-rmse:0.27125	validation_1-rmse:0.31330
[25]	validation_0-rmse:0.26807	validation_1-rmse:0.31087
[26]	validation_0-rmse:0.26695	validation_1-rmse:0.31061
[27]	validation_0-rmse:0.26557	validation_1-rmse:0.30983
[28]	validation_0-rmse:0.26483	validation_1-rmse:0.30959
[29]	validation_0-rmse:0.26218	validation_1-rmse:0.30775
[30]	validation_0-rmse:0.26009	validation_1-rmse:0.30638
[31]	validation_0-rmse:0.25702	validation_1-rmse:0.30447
[32]	validation_0-rmse:0.25219	validation_1-rmse:0.30066
[33]	validation_0-rmse:0.25082	validation_1-rmse:0.30016
[34]	validation_0-rmse:0.25006	validation_1-rmse:0.30001
[35]	validation_0-rmse:0.24881	validation_1-rmse:0.29997
[36]	validation_0-rmse:0.24700	validation_1-rmse:0.29957
[37]	validation_0-rmse:0.24604	validation_1-rmse:0.29944
[38]	validation_0-rmse:0.24537	validation_1-rmse:0.29931
[39]	validation_0-rmse:0.24505	validation_1-rmse:0.29909
[40]	validation_0-rmse:0.24388	validation_1-rmse:0.29895
[41]	validation_0-rmse:0.24261	validation_1-rmse:0.29857

[42]	validation_0-rmse:0.24127	validation_1-rmse:0.29878
[43]	validation_0-rmse:0.23997	validation_1-rmse:0.29840
[44]	validation_0-rmse:0.23906	validation_1-rmse:0.29831
[45]	validation_0-rmse:0.23720	validation_1-rmse:0.29806
[46]	validation_0-rmse:0.23589	validation_1-rmse:0.29711
[47]	validation_0-rmse:0.23477	validation_1-rmse:0.29711
[48]	validation_0-rmse:0.23410	validation_1-rmse:0.29720
[49]	validation_0-rmse:0.23378	validation_1-rmse:0.29726
[50]	validation_0-rmse:0.23307	validation_1-rmse:0.29714
[51]	validation_0-rmse:0.23290	validation_1-rmse:0.29719
[52]	validation_0-rmse:0.23238	validation_1-rmse:0.29728
[53]	validation_0-rmse:0.23174	validation_1-rmse:0.29745
[54]	validation_0-rmse:0.23019	validation_1-rmse:0.29654
[55]	validation_0-rmse:0.22935	validation_1-rmse:0.29631
[56]	validation_0-rmse:0.22672	validation_1-rmse:0.29506
[57]	validation_0-rmse:0.22604	validation_1-rmse:0.29503
[58]	validation_0-rmse:0.22533	validation_1-rmse:0.29528
[59]	validation_0-rmse:0.22463	validation_1-rmse:0.29480
[60]	validation_0-rmse:0.22409	validation_1-rmse:0.29477
[61]	validation_0-rmse:0.22325	validation_1-rmse:0.29470
[62]	validation_0-rmse:0.22239	validation_1-rmse:0.29470
[63]	validation_0-rmse:0.22142	validation_1-rmse:0.29461
[64]	validation_0-rmse:0.22079	validation_1-rmse:0.29465
[65]	validation_0-rmse:0.21931	validation_1-rmse:0.29375
[66]	validation_0-rmse:0.21840	validation_1-rmse:0.29406
[67]	validation_0-rmse:0.21726	validation_1-rmse:0.29411
[68]	validation_0-rmse:0.21638	validation_1-rmse:0.29398
[69]	validation_0-rmse:0.21593	validation_1-rmse:0.29391
[70]	validation_0-rmse:0.21451	validation_1-rmse:0.29330
[71]	validation_0-rmse:0.21333	validation_1-rmse:0.29310
[72]	validation_0-rmse:0.21286	validation_1-rmse:0.29275
[73]	validation_0-rmse:0.21211	validation_1-rmse:0.29260
[74]	validation_0-rmse:0.21102	validation_1-rmse:0.29255
[75]	validation_0-rmse:0.20990	validation_1-rmse:0.29237
[76]	validation_0-rmse:0.20917	validation_1-rmse:0.29246
[77]	validation_0-rmse:0.20814	validation_1-rmse:0.29266
[78]	validation_0-rmse:0.20762	validation_1-rmse:0.29285
[79]	validation_0-rmse:0.20734	validation_1-rmse:0.29269
[80]	validation_0-rmse:0.20727	validation_1-rmse:0.29276
[81]	validation_0-rmse:0.20645	validation_1-rmse:0.29244
[82]	validation_0-rmse:0.20502	validation_1-rmse:0.29156
[83]	validation_0-rmse:0.20409	validation_1-rmse:0.29175

[84]	validation_0-rmse:0.20323	validation_1-rmse:0.29154
[85]	validation_0-rmse:0.20222	validation_1-rmse:0.29140
[86]	validation_0-rmse:0.20213	validation_1-rmse:0.29135
[87]	validation_0-rmse:0.20202	validation_1-rmse:0.29122
[88]	validation_0-rmse:0.20184	validation_1-rmse:0.29115
[89]	validation_0-rmse:0.20122	validation_1-rmse:0.29113
[90]	validation_0-rmse:0.20085	validation_1-rmse:0.29121
[91]	validation_0-rmse:0.20008	validation_1-rmse:0.29090
[92]	validation_0-rmse:0.19988	validation_1-rmse:0.29101
[93]	validation_0-rmse:0.19970	validation_1-rmse:0.29110
[94]	validation_0-rmse:0.19913	validation_1-rmse:0.29120
[95]	validation_0-rmse:0.19850	validation_1-rmse:0.29104
[96]	validation_0-rmse:0.19787	validation_1-rmse:0.29102
[97]	validation_0-rmse:0.19706	validation_1-rmse:0.29103
[98]	validation_0-rmse:0.19631	validation_1-rmse:0.29085
[99]	validation_0-rmse:0.19620	validation_1-rmse:0.29079
[100]	validation_0-rmse:0.19551	validation_1-rmse:0.29071
[101]	validation_0-rmse:0.19493	validation_1-rmse:0.29062
[102]	validation_0-rmse:0.19440	validation_1-rmse:0.29044
[103]	validation_0-rmse:0.19425	validation_1-rmse:0.29045
[104]	validation_0-rmse:0.19393	validation_1-rmse:0.29034
[105]	validation_0-rmse:0.19326	validation_1-rmse:0.29036
[106]	validation_0-rmse:0.19307	validation_1-rmse:0.29041
[107]	validation_0-rmse:0.19251	validation_1-rmse:0.29033
[108]	validation_0-rmse:0.19186	validation_1-rmse:0.29065
[109]	validation_0-rmse:0.19124	validation_1-rmse:0.29053
[110]	validation_0-rmse:0.19090	validation_1-rmse:0.29066
[111]	validation_0-rmse:0.19012	validation_1-rmse:0.29036
[112]	validation_0-rmse:0.18899	validation_1-rmse:0.28999
[113]	validation_0-rmse:0.18818	validation_1-rmse:0.28995
[114]	validation_0-rmse:0.18750	validation_1-rmse:0.28975
[115]	validation_0-rmse:0.18727	validation_1-rmse:0.28975
[116]	validation_0-rmse:0.18677	validation_1-rmse:0.28989
[117]	validation_0-rmse:0.18672	validation_1-rmse:0.28990
[118]	validation_0-rmse:0.18605	validation_1-rmse:0.28986
[119]	validation_0-rmse:0.18557	validation_1-rmse:0.28999
[120]	validation_0-rmse:0.18498	validation_1-rmse:0.28981
[121]	validation_0-rmse:0.18473	validation_1-rmse:0.28989
[122]	validation_0-rmse:0.18469	validation_1-rmse:0.28996
[123]	validation_0-rmse:0.18452	validation_1-rmse:0.28997
[124]	validation_0-rmse:0.18382	validation_1-rmse:0.28962
[125]	validation_0-rmse:0.18352	validation_1-rmse:0.28963

[126]	validation_0-rmse:0.18330	validation_1-rmse:0.28963
[127]	validation_0-rmse:0.18261	validation_1-rmse:0.28935
[128]	validation_0-rmse:0.18253	validation_1-rmse:0.28934
[129]	validation_0-rmse:0.18190	validation_1-rmse:0.28933
[130]	validation_0-rmse:0.18108	validation_1-rmse:0.28946
[131]	validation_0-rmse:0.18010	validation_1-rmse:0.28909
[132]	validation_0-rmse:0.17941	validation_1-rmse:0.28895
[133]	validation_0-rmse:0.17879	validation_1-rmse:0.28896
[134]	validation_0-rmse:0.17859	validation_1-rmse:0.28877
[135]	validation_0-rmse:0.17844	validation_1-rmse:0.28871
[136]	validation_0-rmse:0.17802	validation_1-rmse:0.28904
[137]	validation_0-rmse:0.17735	validation_1-rmse:0.28903
[138]	validation_0-rmse:0.17673	validation_1-rmse:0.28922
[139]	validation_0-rmse:0.17636	validation_1-rmse:0.28915
[140]	validation_0-rmse:0.17586	validation_1-rmse:0.28900
[141]	validation_0-rmse:0.17579	validation_1-rmse:0.28905
[142]	validation_0-rmse:0.17521	validation_1-rmse:0.28906
[143]	validation_0-rmse:0.17483	validation_1-rmse:0.28913
[144]	validation_0-rmse:0.17444	validation_1-rmse:0.28919
[145]	validation_0-rmse:0.17414	validation_1-rmse:0.28917
[146]	validation_0-rmse:0.17365	validation_1-rmse:0.28909
[147]	validation_0-rmse:0.17327	validation_1-rmse:0.28922
[148]	validation_0-rmse:0.17273	validation_1-rmse:0.28917
[149]	validation_0-rmse:0.17217	validation_1-rmse:0.28913

Out[12]:

**XGBRegressor**

```
XGBRegressor(base_score=None, booster=None, callbacks=None,
              colsample_bylevel=None, colsample_bynode=None,
              colsample_bytree=None, early_stopping_rounds=None,
              enable_categorical=False, eval_metric=None, feature_types=None,
              gamma=None, gpu_id=None, grow_policy=None, importance_type=None,
              interaction_constraints=None, learning_rate=None, max_bin=None,
              max_cat_threshold=None, max_cat_to_onehot=None,
              max_delta_step=None, max_depth=5, max_leaves=None,
              min_child_weight=None, missing=nan, monotone_constraints=None,
```

In [13]: `df_train['count'].describe()`

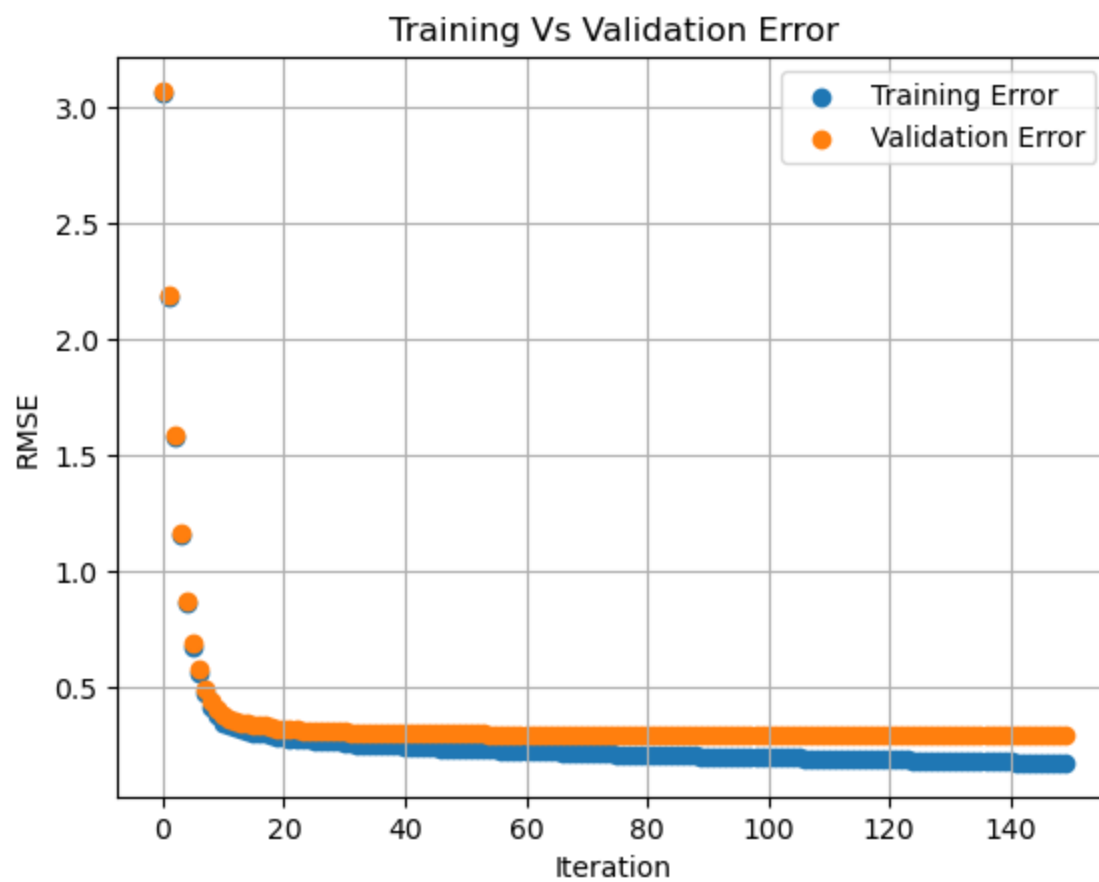
```
Out[13]: count      7620.000000  
         mean        4.583886  
         std         1.429959  
         min         0.693147  
         25%         3.737670  
         50%         4.976734  
         75%         5.652489  
         max         6.885510  
         Name: count, dtype: float64
```

```
In [14]: eval_result = regressor.evals_result()
```

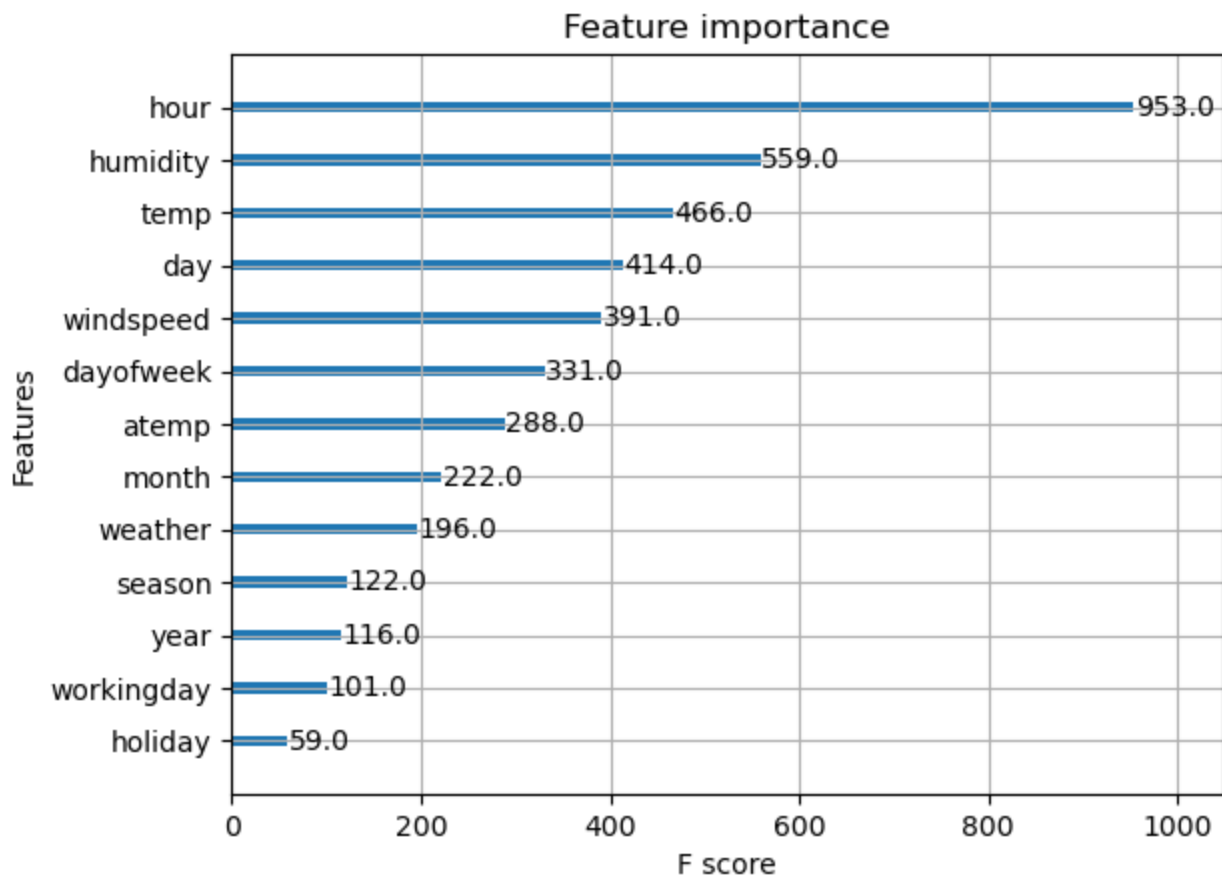
```
In [15]: training_rounds = range(len(eval_result['validation_0']['rmse']))
```

```
In [16]: plt.scatter(x=training_rounds,y=eval_result['validation_0']['rmse'],label='Training Error')  
         plt.scatter(x=training_rounds,y=eval_result['validation_1']['rmse'],label='Validation Error')  
         plt.grid(True)  
         plt.xlabel('Iteration')  
         plt.ylabel('RMSE')  
         plt.title('Training Vs Validation Error')  
         plt.legend()  
         plt.show()
```





```
In [17]: xgb.plot_importance(regressor)
plt.show()
```



```
In [18]: # Updated - Changed to validation dataset
# Compare actual vs predicted performance with dataset not seen by the model before
df = pd.read_csv(validation_file,names=columns)
```

```
In [19]: df.head()
```

```
Out[19]:
```

	count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
0	6.095825	3	0	1	2	28.70	33.335	79	12.9980	2011	7	7	3	8
1	5.961005	2	0	0	1	32.80	37.880	55	12.9980	2011	6	11	5	13
2	1.098612	1	0	1	1	14.76	16.665	40	19.9995	2011	2	14	0	2
3	3.891820	1	0	1	1	9.02	9.090	47	36.9974	2011	2	8	1	10
4	4.025352	4	0	0	1	10.66	15.150	87	0.0000	2011	12	4	6	8

```
In [20]: X_test = df.iloc[:,1:]
print(X_test[:5])
```

	season	holiday	workingday	weather	temp	atemp	humidity	windspeed
0	3	0	1	2	28.70	33.335	79	12.9980 \
1	2	0	0	1	32.80	37.880	55	12.9980
2	1	0	1	1	14.76	16.665	40	19.9995
3	1	0	1	1	9.02	9.090	47	36.9974
4	4	0	0	1	10.66	15.150	87	0.0000

	year	month	day	dayofweek	hour
0	2011	7	7	3	8
1	2011	6	11	5	13
2	2011	2	14	0	2
3	2011	2	8	1	10
4	2011	12	4	6	8

```
In [21]: result = regressor.predict(X_test)
```

```
In [22]: result[:5]
```

```
Out[22]: array([6.259567 , 5.966802 , 1.3931606, 3.9536645, 4.000689 ],
              dtype=float32)
```

```
In [23]: df.head()
```

Out[23]:

	count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
0	6.095825	3	0	1	2	28.70	33.335	79	12.9980	2011	7	7	3	8
1	5.961005	2	0	0	1	32.80	37.880	55	12.9980	2011	6	11	5	13
2	1.098612	1	0	1	1	14.76	16.665	40	19.9995	2011	2	14	0	2
3	3.891820	1	0	1	1	9.02	9.090	47	36.9974	2011	2	8	1	10
4	4.025352	4	0	0	1	10.66	15.150	87	0.0000	2011	12	4	6	8

In [24]: `df['count_predicted'] = result`In [25]: `df.head()`

Out[25]:

	count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour	count_pre
0	6.095825	3	0	1	2	28.70	33.335	79	12.9980	2011	7	7	3	8	6.2
1	5.961005	2	0	0	1	32.80	37.880	55	12.9980	2011	6	11	5	13	5.9
2	1.098612	1	0	1	1	14.76	16.665	40	19.9995	2011	2	14	0	2	1.3
3	3.891820	1	0	1	1	9.02	9.090	47	36.9974	2011	2	8	1	10	3.9
4	4.025352	4	0	0	1	10.66	15.150	87	0.0000	2011	12	4	6	8	4.0

In [26]: `# Negative Values are #DWB# NOT!! #DWB# predicted`  
`df['count_predicted'].describe()`

Out[26]:

count	3266.000000
mean	4.592924
std	1.380849
min	0.626428
25%	3.790787
50%	4.991671
75%	5.625294
max	6.830800

Name: count\_predicted, dtype: float64

```
In [27]: df[df['count_predicted'] < 0]
```

```
Out[27]:
```

count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour	count_predicted
-------	--------	---------	------------	---------	------	-------	----------	-----------	------	-------	-----	-----------	------	-----------------



```
In [28]: def adjust_count(x):  
        if x < 0:  
            return 0  
        else:  
            return x
```

```
In [29]: df['count_predicted'] = df['count_predicted'].map(adjust_count)
```

```
In [30]: df[df['count_predicted'] < 0]
```

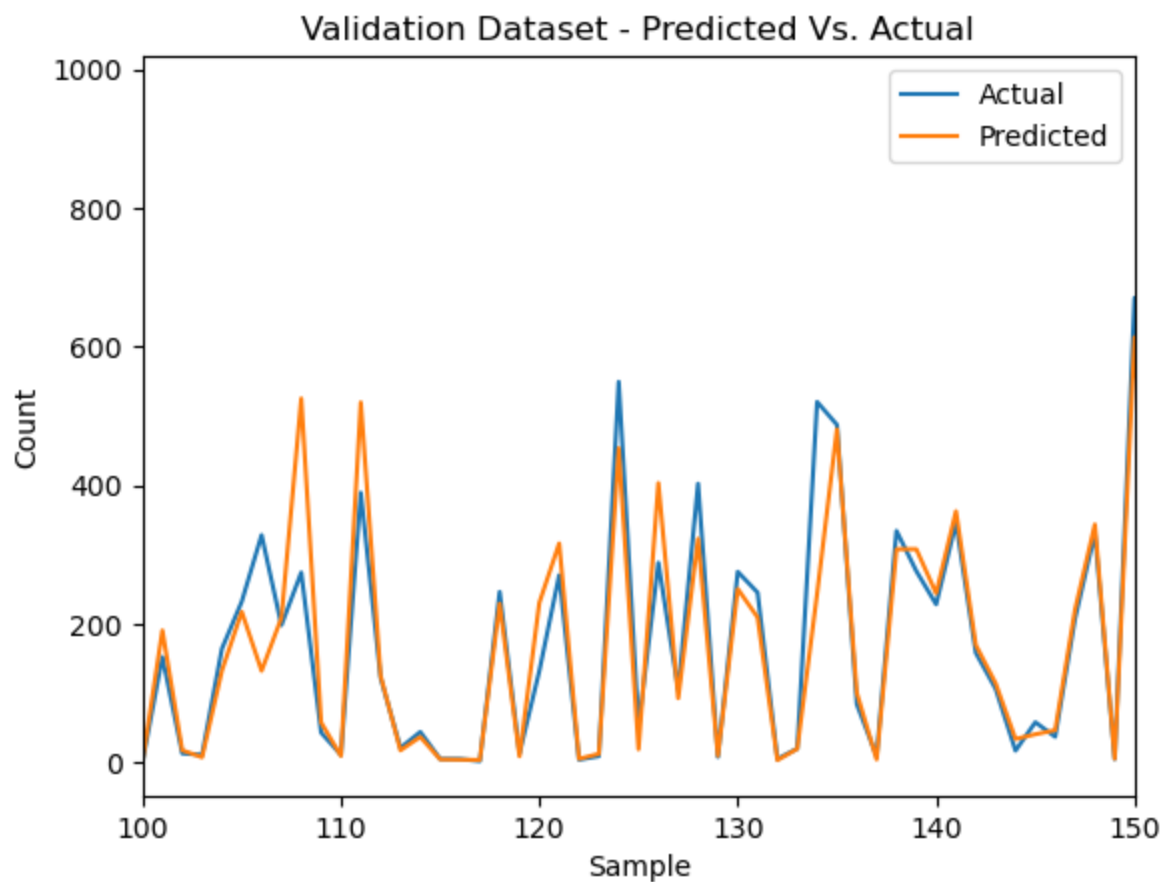
```
Out[30]:
```

count	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour	count_predicted
-------	--------	---------	------------	---------	------	-------	----------	-----------	------	-------	-----	-----------	------	-----------------



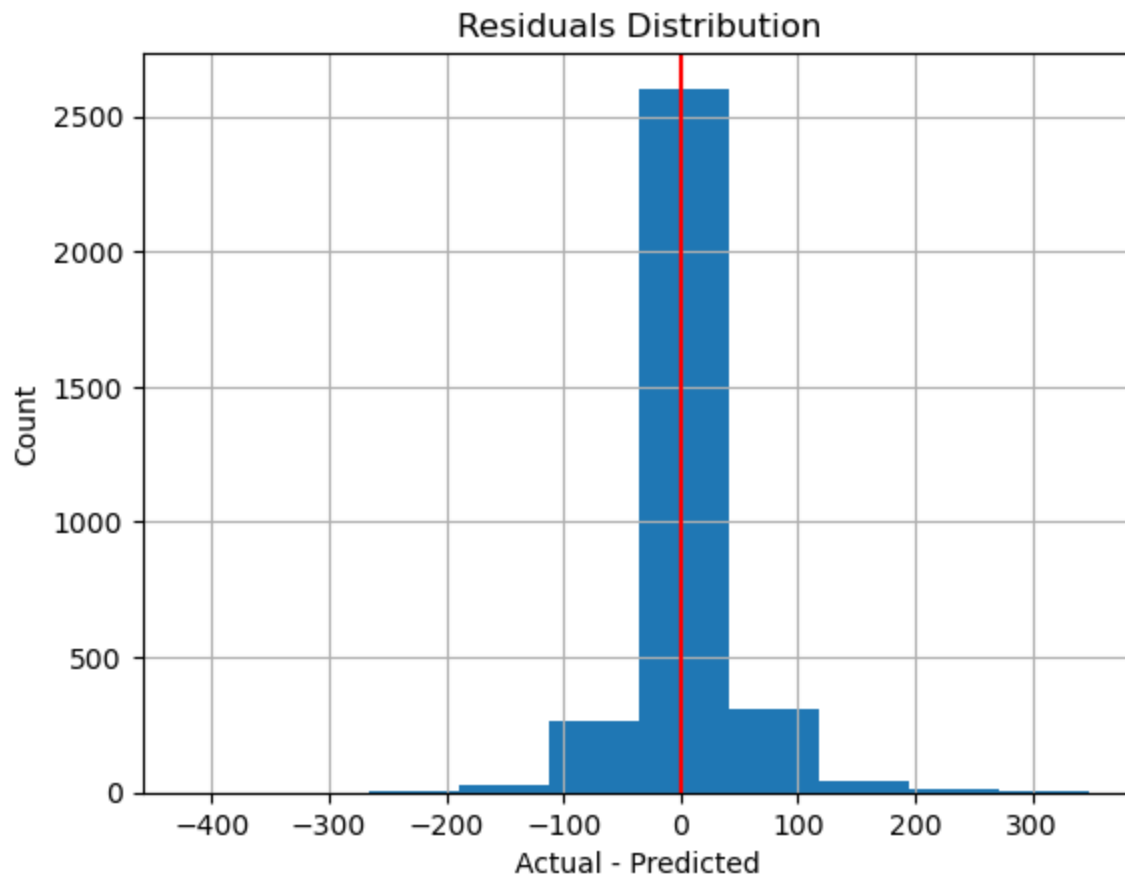
```
In [31]: df['count'] = df['count'].map(np.expm1)  
df['count_predicted'] = df['count_predicted'].map(np.expm1)
```

```
In [32]: # Actual Vs Predicted  
plt.plot(df['count'], label='Actual')  
plt.plot(df['count_predicted'], label='Predicted')  
plt.xlabel('Sample')  
plt.ylabel('Count')  
plt.xlim([100, 150])  
plt.title('Validation Dataset - Predicted Vs. Actual')  
plt.legend()  
plt.show()
```



```
In [33]: # Over prediction and Under Prediction needs to be balanced
# Training Data Residuals
residuals = (df['count'] - df['count_predicted'])

plt.hist(residuals)
plt.grid(True)
plt.xlabel('Actual - Predicted')
plt.ylabel('Count')
plt.title('Residuals Distribution')
plt.axvline(color='r')
plt.show()
```



```
In [34]: value_counts = (residuals > 0).value_counts(sort=False)
print(' Under Estimation: {0:.2f}'.format(value_counts[True]/len(residuals)))
print(' Over Estimation: {0:.2f}'.format(value_counts[False]/len(residuals)))
```

```
Under Estimation: 0.53
Over Estimation: 0.47
```

```
In [35]: import sklearn.metrics as metrics
print("RMSE: {0:.2f}".format(metrics.mean_squared_error(df['count'],
                                                         df['count_predicted'])**.5))
```

```
RMSE: 42.41
```

```
In [36]: # Metric Use By Kaggle
def compute_rmsle(y_true, y_pred):
```

```

if type(y_true) != np.ndarray:
    y_true = np.array(y_true)

if type(y_pred) != np.ndarray:
    y_pred = np.array(y_pred)

return(np.average((np.log1p(y_pred) - np.log1p(y_true))**2)**.5)

```

```
In [37]: print("RMSLE: {0:.2f}".format(compute_rmsle(df['count'],df['count_predicted'])))
```

RMSLE: 0.29

```
In [38]: # Prepare Data for Submission to Kaggle
df_test = pd.read_csv(test_file,parse_dates=['datetime'])
```

```
In [39]: df_test.head()
```

```
Out[39]:
```

	datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
<b>0</b>	2011-01-20 00:00:00	1	0	1	1	10.66	11.365	56	26.0027	2011	1	20	3	0
<b>1</b>	2011-01-20 01:00:00	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	1
<b>2</b>	2011-01-20 02:00:00	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	2
<b>3</b>	2011-01-20 03:00:00	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	3
<b>4</b>	2011-01-20 04:00:00	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	4

```
In [40]: X_test = df_test.iloc[:,1:] # Exclude datetime for prediction
```

```
In [41]: X_test.head()
```



```
Out[41]:
```

	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour
0	1	0	1	1	10.66	11.365	56	26.0027	2011	1	20	3	0
1	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	1
2	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	2
3	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	3
4	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	4

```
In [42]: result = regressor.predict(X_test)
```

```
In [43]: result[:5]
```

```
Out[43]: array([2.0552852, 1.5469345, 1.3397862, 1.1393155, 1.051079 ],
              dtype=float32)
```

```
In [44]: np.exp1(result)
```

```
Out[44]: array([ 6.809065 ,  3.6970491,  2.818227 , ..., 125.76962 ,
                86.26166 , 44.861454 ], dtype=float32)
```

```
In [45]: # Convert result to actual count
df_test["count"] = np.exp1(result)
```

```
In [46]: df_test.head()
```

Out[46]:

	datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour	count
0	2011-01-20 00:00:00	1	0	1	1	10.66	11.365	56	26.0027	2011	1	20	3	0	6.809065
1	2011-01-20 01:00:00	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	1	3.697049
2	2011-01-20 02:00:00	1	0	1	1	10.66	13.635	56	0.0000	2011	1	20	3	2	2.818227
3	2011-01-20 03:00:00	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	3	2.124629
4	2011-01-20 04:00:00	1	0	1	1	10.66	12.880	56	11.0014	2011	1	20	3	4	1.860736

In [47]: `df_test[df_test["count"] < 0]`

Out[47]:

	datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	year	month	day	dayofweek	hour	count
--	----------	--------	---------	------------	---------	------	-------	----------	-----------	------	-------	-----	-----------	------	-------

In [48]: `df_test[['datetime','count']].to_csv('predicted_count.csv',index=False)`

In [49]:

```
# RMSLE (Kaggle) Score
# Test 1: 0.62 # Chandra
#DWB# Test 1: 0.71796
# Test 2(log of count): 0.40 # Chandra
#DWB# Test2 (log of count): 0.42225
```

In [ ]: