

```
In [15]: ##### import libraries
import joblib
import pandas as pd
```

```
In [16]: #####Loading the training model
file_name = 'RANDOM_FOREST_MODEL.SAV'
my_model = joblib.load(file_name)
```

```
In [17]: ##### Loading the test_data
Test_data = pd.read_csv('Downloads/test.csv')
Test_data.head()
```

Out[17]:

| | datetime | season | holiday | workingday | weather | temp | atemp | humidity | windspeed | casual | registered |
|---|--------------------|--------|---------|------------|---------|-------|--------|----------|-----------|--------|------------|
| 0 | 2012-06-30 1:00:00 | 3 | 0 | 0 | 3 | 26.24 | 28.790 | 89.0 | 15.0013 | 3 | 55 |
| 1 | 2012-06-30 2:00:00 | 3 | 0 | 0 | 2 | 26.24 | 28.790 | 89.0 | 0.0000 | 7 | 54 |
| 2 | 2012-06-30 3:00:00 | 3 | 0 | 0 | 2 | 26.24 | 28.790 | 89.0 | 0.0000 | 3 | 20 |
| 3 | 2012-06-30 4:00:00 | 3 | 0 | 0 | 2 | 25.42 | 27.275 | 94.0 | 0.0000 | 3 | 15 |
| 4 | 2012-06-30 5:00:00 | 3 | 0 | 0 | 1 | 26.24 | 28.790 | 89.0 | 11.0014 | 3 | 7 |

```
In [18]: ##### Doing the data preprocessing for testing_data as same as previously did for the traing_data.
Test_data['datetime'] = pd.to_datetime(Test_data['datetime'])
Test_data['dayofweek'] = Test_data['datetime'].dt.dayofweek
Test_data['month'] = Test_data['datetime'].dt.month
Test_data['year'] = Test_data['datetime'].dt.year
Test_data['hour'] = Test_data['datetime'].dt.hour
Test_data.head()
```

Out[18]:

| | datetime | season | holiday | workingday | weather | temp | atemp | humidity | windspeed | casual | registered | dayofweek | month | year | hour |
|---|---------------------|--------|---------|------------|---------|-------|--------|----------|-----------|--------|------------|-----------|-------|------|------|
| 0 | 2012-06-30 01:00:00 | 3 | 0 | 0 | 3 | 26.24 | 28.790 | 89.0 | 15.0013 | 3 | 55 | 5 | 6 | 2012 | 1 |
| 1 | 2012-06-30 02:00:00 | 3 | 0 | 0 | 2 | 26.24 | 28.790 | 89.0 | 0.0000 | 7 | 54 | 5 | 6 | 2012 | 2 |
| 2 | 2012-06-30 03:00:00 | 3 | 0 | 0 | 2 | 26.24 | 28.790 | 89.0 | 0.0000 | 3 | 20 | 5 | 6 | 2012 | 3 |
| 3 | 2012-06-30 04:00:00 | 3 | 0 | 0 | 2 | 25.42 | 27.275 | 94.0 | 0.0000 | 3 | 15 | 5 | 6 | 2012 | 4 |
| 4 | 2012-06-30 05:00:00 | 3 | 0 | 0 | 1 | 26.24 | 28.790 | 89.0 | 11.0014 | 3 | 7 | 5 | 6 | 2012 | 5 |

```
In [19]: ##### dropping the unnecessary features.
Test_data = Test_data.drop(['datetime', 'atemp'], axis=1)
Test_data.head()
```

Out[19]:

| | season | holiday | workingday | weather | temp | humidity | windspeed | casual | registered | dayofweek | month | year | hour |
|---|--------|---------|------------|---------|-------|----------|-----------|--------|------------|-----------|-------|------|------|
| 0 | 3 | 0 | 0 | 3 | 26.24 | 89.0 | 15.0013 | 3 | 55 | 5 | 6 | 2012 | 1 |
| 1 | 3 | 0 | 0 | 2 | 26.24 | 89.0 | 0.0000 | 7 | 54 | 5 | 6 | 2012 | 2 |
| 2 | 3 | 0 | 0 | 2 | 26.24 | 89.0 | 0.0000 | 3 | 20 | 5 | 6 | 2012 | 3 |
| 3 | 3 | 0 | 0 | 2 | 25.42 | 94.0 | 0.0000 | 3 | 15 | 5 | 6 | 2012 | 4 |
| 4 | 3 | 0 | 0 | 1 | 26.24 | 89.0 | 11.0014 | 3 | 7 | 5 | 6 | 2012 | 5 |

```
In [22]: ##### Now using the trained model and using the test_data for prediction.
##### printing the predictions
model_prediction = my_model.predict(Test_data)
print(model_prediction)

[57.96 60.7  23.02 ... 89.6  61.05 49.19]
```

```
In [23]: ##### Now predicted the counts--('no of total rentals').
### Adding the prediction values as a predicted column to the test_data to compare the actual and predicted values.
Test_data['predicted'] = model_prediction
```

```
In [24]: Test_data
```

Out[24]:

| | season | holiday | workingday | weather | temp | humidity | windspeed | casual | registered | dayofweek | month | year | hour | predicted |
|------|--------|---------|------------|---------|-------|----------|-----------|--------|------------|-----------|-------|------|------|-----------|
| 0 | 3 | 0 | 0 | 3 | 26.24 | 89.0 | 15.0013 | 3 | 55 | 5 | 6 | 2012 | 1 | 57.96 |
| 1 | 3 | 0 | 0 | 2 | 26.24 | 89.0 | 0.0000 | 7 | 54 | 5 | 6 | 2012 | 2 | 60.70 |
| 2 | 3 | 0 | 0 | 2 | 26.24 | 89.0 | 0.0000 | 3 | 20 | 5 | 6 | 2012 | 3 | 23.02 |
| 3 | 3 | 0 | 0 | 2 | 25.42 | 94.0 | 0.0000 | 3 | 15 | 5 | 6 | 2012 | 4 | 18.01 |
| 4 | 3 | 0 | 0 | 1 | 26.24 | 89.0 | 11.0014 | 3 | 7 | 5 | 6 | 2012 | 5 | 10.01 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 4394 | 1 | 0 | 1 | 2 | 10.66 | 60.0 | 11.0014 | 11 | 108 | 0 | 12 | 2012 | 19 | 119.10 |
| 4395 | 1 | 0 | 1 | 2 | 10.66 | 60.0 | 11.0014 | 8 | 81 | 0 | 12 | 2012 | 20 | 89.05 |
| 4396 | 1 | 0 | 1 | 1 | 10.66 | 60.0 | 11.0014 | 7 | 83 | 0 | 12 | 2012 | 21 | 89.60 |
| 4397 | 1 | 0 | 1 | 1 | 10.66 | 56.0 | 8.9981 | 13 | 48 | 0 | 12 | 2012 | 22 | 61.05 |
| 4398 | 1 | 0 | 1 | 1 | 10.66 | 65.0 | 8.9981 | 12 | 37 | 0 | 12 | 2012 | 23 | 49.19 |

4399 rows × 14 columns

```
In [28]: ##### Converting the final_data into csv(using "to_csv") file by giving the path .
csv_file_path = 'downloads/final_data.csv'
final_data.to_csv(csv_file_path,index=False)
print(f'Data has been saved to {csv_file_path}')
```

Data has been saved to downloads/final_data.csv

```
In [29]: ##### About Data -
#####Bike sharing systems are a means of renting bicycles where the process of obtaining membership,
#####rental, and bike return is automated via a network of kiosk locations throughout a city.
#####Using these systems, people are able to rent a bike from one location and return it to different
#####place on an as-needed basis. Currently, there are over 500 bike-sharing programs around the world.

#####The data generated by these systems makes them attractive for researchers because the duration of
#####travel, departure location, arrivallocation, and time elapsed is explicitly recorded.
#####Bike sharing systems therefore function as a sensor network, which can be used for studying mobility
#####in a city.

##### USED THE FOLLOWING DATA-SETS:
#####      1.Train.csv : Use this dataset to train the model. This file contains all the weather related
#####      features as well as the target variable “count”. Train dataset is comprised of first
#####      18 months.

###      2. test.csv : Use the trained model to predict the count of total rentals for each hour during the
###      next 6 months.

##### USING THE RANDOM_FOREST_REGRESSOR CREATED THE MACHINE_LEARNING_MODEL
##### THE PERFORMANCE OF THE MODEL IS EVALUATED BY THE 1.' MEAN SQUARE ERROR',2.R_2 SCORE,3.'ROOT MEAN SQUARED LOG ERROR'
##### USING THE RANDOM_FOREST_REGRESSOR THE MODEL HAS PERFORMED. 1.MEAN SQUARE ERROR = (16.9846).
#####      2.R_2 SCORE = (0.9993).
#####      3.ROOT MEAN SQUARED LOG ERROR = (0.01119)
```

```
In [ ]:
```