

```
In [52]: import pandas as pd
import numpy as np
```

```
In [53]: df = pd.read_csv('webtraffic.csv')
```

```
In [54]: df.head()
```

```
Out[54]:
```

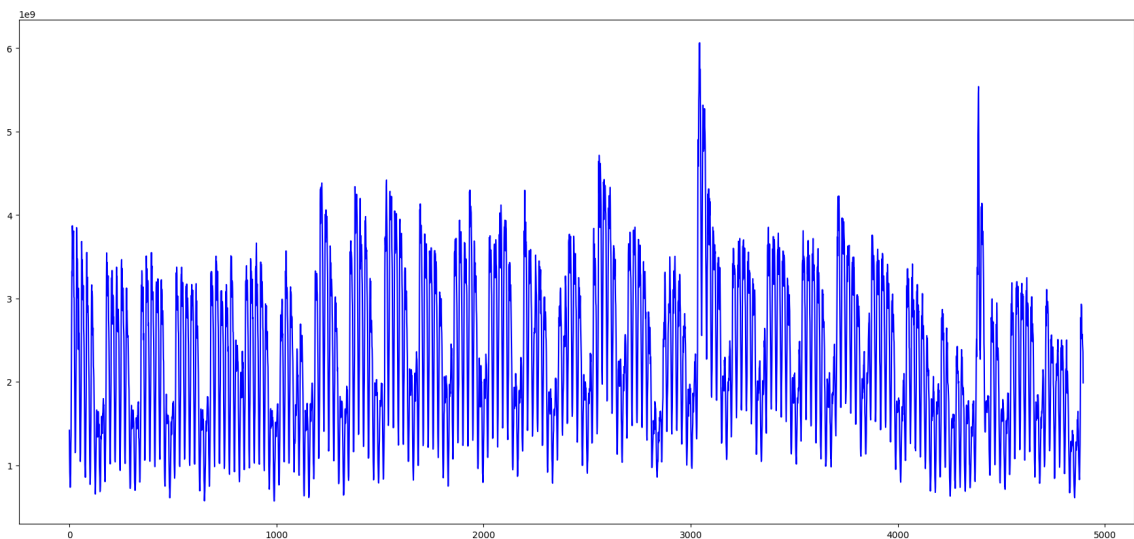
	Hour Index	Sessions
0	0	1418159421
1	1	1113769116
2	2	919158921
3	3	822352824
4	4	735526737

```
In [55]: df.shape
```

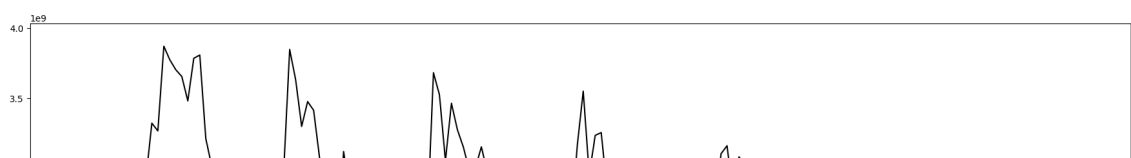
```
Out[55]: (4896, 2)
```

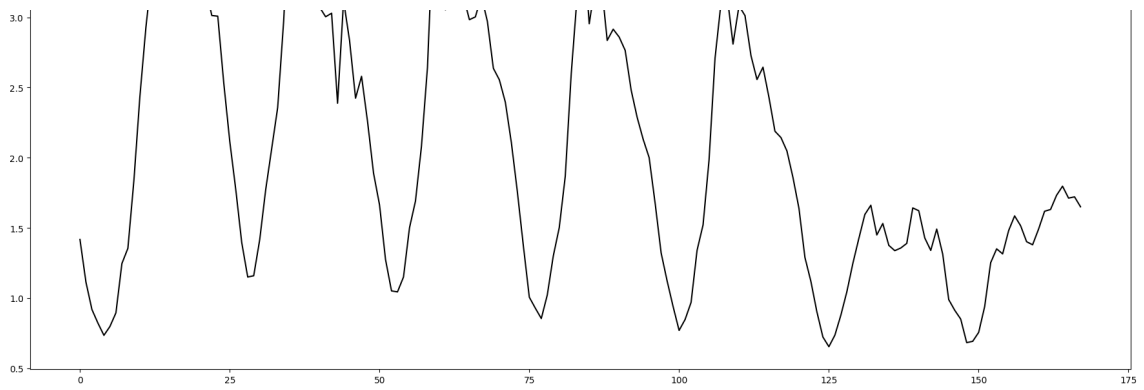
Data Exploration for Web Traffic Forecasting

```
In [56]: import matplotlib.pyplot as plt
sessions = df['Sessions'].values
ar = np.arange(len(sessions))
plt.figure(figsize=(22,10))
plt.plot(ar, sessions, 'b')
plt.show()
```



```
In [57]: #first week web traffic
sample = sessions[:168]
ar = np.arange(len(sample))
plt.figure(figsize=(22,10))
plt.plot(ar, sample, 'black')
plt.show()
```





Data Preparation for Web Traffic Forecasting

```
In [58]: def prepare_data(seq,num):
          x=[]
          y=[]
          for i in range(0,(len(seq)-num),1):

              input_ = seq[i:i+num]
              output = seq[i+num]

              x.append(input_)
              y.append(output)

          return np.array(x), np.array(y)
```

```
In [59]: num=168 #1 week = 168hrs
          x,y= prepare_data(sessions,num)
          print(len(x))
```

4728

Split the Dataset

```
In [60]: ind = int(0.9 * len(x)) #split into 90:10
          x_tr = x[:ind]
          y_tr = y[:ind]
          x_val=x[ind:]
          y_val=y[ind:]
```

```
In [61]: from sklearn.preprocessing import StandardScaler
          #normalize the inputs
          x_scaler= StandardScaler()
          x_tr = x_scaler.fit_transform(x_tr)
          x_val= x_scaler.transform(x_val)
          #reshaping the output for normalization
          y_tr=y_tr.reshape(len(y_tr),1)
          y_val=y_val.reshape(len(y_val),1)
          #normalize the output
          y_scaler=StandardScaler()
          y_tr = y_scaler.fit_transform(y_tr)[:,:0]
          y_val = y_scaler.transform(y_val)[:,:0]
```

```
In [62]: #reshaping input data
          x_tr= x_tr.reshape(x_tr.shape[0],x_tr.shape[1],1)
          x_val= x_val.reshape(x_val.shape[0],x_val.shape[1],1)
          print(x_tr.shape)
```

(4255, 168, 1)

Model Building for Web Traffic Forecasting

```
In [63]: from keras.models import *
```

```

from keras.layers import *
from keras.callbacks import *
from tensorflow import keras
# define model
model = Sequential()
model.add(LSTM(128,input_shape=(168,1)))
model.add(Dense(64,activation='relu'))
model.add(Dense(1,activation='linear'))

```

In [64]: `model.summary()`

Model: "sequential_2"

Layer (type)	Output Shape	Param #
=====		
lstm_1 (LSTM)	(None, 128)	66560
dense_4 (Dense)	(None, 64)	8256
dense_5 (Dense)	(None, 1)	65
=====		
Total params: 74,881		
Trainable params: 74,881		
Non-trainable params: 0		

In [65]:

```

# Define the optimizer and loss
model.compile(loss='mse',optimizer='adam')
#Define the callback to save the best model during the training
mc = ModelCheckpoint('best_model.hdf5', monitor='val_loss',
    verbose=1, save_best_only=True, mode='min')
# Train the model for 30 epochs with batch size of 32:
history=model.fit(x_tr, y_tr ,epochs=30, batch_size=32,
    validation_data=(x_val,y_val), callbacks=[mc])

```

```

Epoch 1/30
133/133 [=====] - ETA: 0s - loss: 0.1544
Epoch 1: val_loss improved from inf to 0.03151, saving model to best_model.hdf5
133/133 [=====] - 32s 191ms/step - loss: 0.1544 - val_loss: 0.0315
Epoch 2/30
133/133 [=====] - ETA: 0s - loss: 0.0399
Epoch 2: val_loss did not improve from 0.03151
133/133 [=====] - 19s 141ms/step - loss: 0.0399 - val_loss: 0.0362
Epoch 3/30
133/133 [=====] - ETA: 0s - loss: 0.0350
Epoch 3: val_loss improved from 0.03151 to 0.03004, saving model to best_model.hdf5
133/133 [=====] - 18s 136ms/step - loss: 0.0350 - val_loss: 0.0300
Epoch 4/30
133/133 [=====] - ETA: 0s - loss: 0.0332
Epoch 4: val_loss improved from 0.03004 to 0.02779, saving model to best_model.hdf5
133/133 [=====] - 24s 182ms/step - loss: 0.0332 - val_loss: 0.0278
Epoch 5/30
133/133 [=====] - ETA: 0s - loss: 0.0328
Epoch 5: val_loss improved from 0.02779 to 0.02505, saving model to best_model.hdf5
133/133 [=====] - 31s 233ms/step - loss: 0.0328 - val_loss: 0.0250
Epoch 6/30
133/133 [=====] - ETA: 0s - loss: 0.0302
Epoch 6: val_loss did not improve from 0.02505
133/133 [=====] - 25s 189ms/step - loss: 0.0302 - val_loss: 0.0272
Epoch 7/30
133/133 [=====] - ETA: 0s - loss: 0.0274
Epoch 7: val_loss improved from 0.02505 to 0.02405, saving model to best_model.hdf5
133/133 [=====] - 24s 183ms/step - loss: 0.0274 - val_loss: 0.0241
Epoch 8/30

```

```
Epoch 8/30
133/133 [=====] - ETA: 0s - loss: 0.0276
Epoch 8: val_loss improved from 0.02405 to 0.02384, saving model to best_model.hdf5
133/133 [=====] - 25s 189ms/step - loss: 0.0276 - val_loss: 0.0238
Epoch 9/30
133/133 [=====] - ETA: 0s - loss: 0.0261
Epoch 9: val_loss did not improve from 0.02384
133/133 [=====] - 24s 179ms/step - loss: 0.0261 - val_loss: 0.0255
Epoch 10/30
133/133 [=====] - ETA: 0s - loss: 0.0248
Epoch 10: val_loss did not improve from 0.02384
133/133 [=====] - 31s 233ms/step - loss: 0.0248 - val_loss: 0.0245
Epoch 11/30
133/133 [=====] - ETA: 0s - loss: 0.0217
Epoch 11: val_loss did not improve from 0.02384
133/133 [=====] - 27s 203ms/step - loss: 0.0217 - val_loss: 0.0251
Epoch 12/30
133/133 [=====] - ETA: 0s - loss: 0.0200
Epoch 12: val_loss improved from 0.02384 to 0.01826, saving model to best_model.hdf5
133/133 [=====] - 23s 176ms/step - loss: 0.0200 - val_loss: 0.0183
Epoch 13/30
133/133 [=====] - ETA: 0s - loss: 0.0201
Epoch 13: val_loss improved from 0.01826 to 0.01791, saving model to best_model.hdf5
133/133 [=====] - 23s 174ms/step - loss: 0.0201 - val_loss: 0.0179
Epoch 14/30
133/133 [=====] - ETA: 0s - loss: 0.0191
Epoch 14: val_loss did not improve from 0.01791
133/133 [=====] - 21s 160ms/step - loss: 0.0191 - val_loss: 0.0199
Epoch 15/30
133/133 [=====] - ETA: 0s - loss: 0.0186
Epoch 15: val_loss improved from 0.01791 to 0.01669, saving model to best_model.hdf5
133/133 [=====] - 21s 155ms/step - loss: 0.0186 - val_loss: 0.0167
Epoch 16/30
133/133 [=====] - ETA: 0s - loss: 0.0174
Epoch 16: val_loss improved from 0.01669 to 0.01448, saving model to best_model.hdf5
133/133 [=====] - 20s 151ms/step - loss: 0.0174 - val_loss: 0.0145
Epoch 17/30
133/133 [=====] - ETA: 0s - loss: 0.0182
Epoch 17: val_loss did not improve from 0.01448
133/133 [=====] - 19s 146ms/step - loss: 0.0182 - val_loss: 0.0172
Epoch 18/30
133/133 [=====] - ETA: 0s - loss: 0.0176
Epoch 18: val_loss did not improve from 0.01448
133/133 [=====] - 20s 152ms/step - loss: 0.0176 - val_loss: 0.0190
Epoch 19/30
133/133 [=====] - ETA: 0s - loss: 0.0163
Epoch 19: val_loss did not improve from 0.01448
133/133 [=====] - 20s 149ms/step - loss: 0.0163 - val_loss: 0.0160
Epoch 20/30
133/133 [=====] - ETA: 0s - loss: 0.0167
Epoch 20: val_loss did not improve from 0.01448
133/133 [=====] - 26s 199ms/step - loss: 0.0167 - val_loss: 0.0166
Epoch 21/30
133/133 [=====] - ETA: 0s - loss: 0.0161
Epoch 21: val_loss did not improve from 0.01448
133/133 [=====] - 27s 204ms/step - loss: 0.0161 - val_loss: 0.0168
Epoch 22/30
133/133 [=====] - ETA: 0s - loss: 0.0164
Epoch 22: val_loss did not improve from 0.01448
133/133 [=====] - 19s 145ms/step - loss: 0.0164 - val_loss: 0.0165
```

```

Epoch 23/30
133/133 [=====] - ETA: 0s - loss: 0.0164
Epoch 23: val_loss did not improve from 0.01448
133/133 [=====] - 19s 144ms/step - loss: 0.0164 - val_loss: 0.0192
Epoch 24/30
133/133 [=====] - ETA: 0s - loss: 0.0162
Epoch 24: val_loss did not improve from 0.01448
133/133 [=====] - 19s 145ms/step - loss: 0.0162 - val_loss: 0.0230
Epoch 25/30
133/133 [=====] - ETA: 0s - loss: 0.0180
Epoch 25: val_loss did not improve from 0.01448
133/133 [=====] - 19s 143ms/step - loss: 0.0180 - val_loss: 0.0166
Epoch 26/30
133/133 [=====] - ETA: 0s - loss: 0.0155
Epoch 26: val_loss did not improve from 0.01448
133/133 [=====] - 19s 144ms/step - loss: 0.0155 - val_loss: 0.0159
Epoch 27/30
133/133 [=====] - ETA: 0s - loss: 0.0158
Epoch 27: val_loss did not improve from 0.01448
133/133 [=====] - 19s 146ms/step - loss: 0.0158 - val_loss: 0.0188
Epoch 28/30
133/133 [=====] - ETA: 0s - loss: 0.0153
Epoch 28: val_loss did not improve from 0.01448
133/133 [=====] - 19s 144ms/step - loss: 0.0153 - val_loss: 0.0154
Epoch 29/30
133/133 [=====] - ETA: 0s - loss: 0.0152
Epoch 29: val_loss improved from 0.01448 to 0.01438, saving model to best_model.hdf5
133/133 [=====] - 19s 145ms/step - loss: 0.0152 - val_loss: 0.0144
Epoch 30/30
133/133 [=====] - ETA: 0s - loss: 0.0160
Epoch 30: val_loss did not improve from 0.01438
133/133 [=====] - 18s 138ms/step - loss: 0.0160 - val_loss: 0.0201

```

```
In [66]: model.load_weights('best_model.hdf5')
```

```
In [67]: mse = model.evaluate(x_val,y_val)
print("Mean Square Error:",mse)
```

```

4/15 [=====>.....] - ETA: 0s - loss: 0.018515/15 [=====]
=====] - 1s 64ms/step - loss: 0.0144
Mean Square Error: 0.014384998008608818

```

Baseline Model with Forecasting

```
In [68]: # build a simple moving average model
def compute_moving_average(data):
    pred=[]
    for i in data:
        avg=np.sum(i)/len(i)
        pred.append(avg)
    return np.array(pred)
# reshape the data
x_resaped = x_val.reshape(-1,168)
# get predictions
y_pred = compute_moving_average(x_resaped)
# evaluate the performance of model on the validation data
mse = np.sum ( (y_val - y_pred) **2 ) / (len(y_val))
print("Mean square of error:- ",mse)
```

Mean square of error:- 0.5546025834434455

Web Traffic Forecasting

```
In [69]: def forecast(x_val, no_of_pred, ind):
          predictions=[]
          #initialize the array with a weeks data
          temp=x_val[ind]
          for i in range(no_of_pred):
              #predict for the next hour
              pred=model.predict(temp.reshape(1,-1,1))[0][0]

              #append the prediction as the last element of array
              temp = np.insert(temp,len(temp),pred)
              predictions.append(pred)
              #ignore the first element of array
              temp = temp[1:]
          return predictions
```

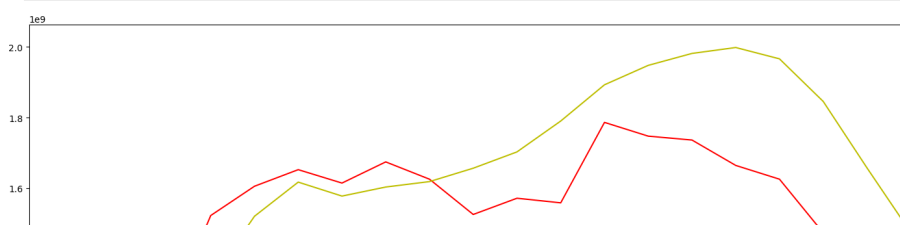
```
In [70]: no_of_pred =24
          ind=72
          y_pred= forecast(x_val,no_of_pred,ind)
          y_true = y_val[ind:ind+(no_of_pred)]
```

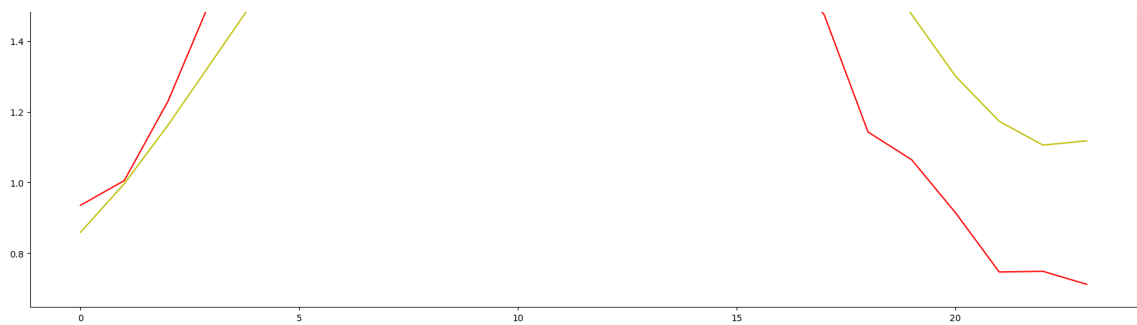
```
1/1 [=====] - 1s 920ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 40ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 26ms/step
1/1 [=====] - 0s 33ms/step
1/1 [=====] - 0s 30ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 24ms/step
1/1 [=====] - 0s 31ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 24ms/step
1/1 [=====] - 0s 33ms/step
1/1 [=====] - 0s 56ms/step
1/1 [=====] - 0s 24ms/step
1/1 [=====] - 0s 33ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 32ms/step
```

```
In [71]: y_true = np.array(y_true)
          y_true = y_true.reshape(-1, 1)

          y_pred = np.array(y_pred)
          y_pred = y_pred.reshape(-1, 1)
          y_true= y_scaler.inverse_transform(y_true)
          y_pred= y_scaler.inverse_transform(y_pred)
```

```
In [72]: def plot(y_true,y_pred):
          ar = np.arange(len(y_true))
          plt.figure(figsize=(22,10))
          plt.plot(ar, y_true,'r')
          plt.plot(ar, y_pred,'y')
          plt.show()
          plot(y_true,y_pred)
```





CNN Model with Forecasting

```
In [73]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import *
from tensorflow.keras.callbacks import *
model= Sequential()
model.add(Conv1D(64, 3, padding='same', activation='relu',input_shape=(num,1)))
model.add(Conv1D(32, 5, padding='same', activation='relu',input_shape=(num,1)))
model.add(Flatten())
model.add(Dense(64,activation='relu'))
model.add(Dense(1,activation='linear'))
model.summary()
```

Model: "sequential_3"

Layer (type)	Output Shape	Param #
conv1d_2 (Conv1D)	(None, 168, 64)	256
conv1d_3 (Conv1D)	(None, 168, 32)	10272
flatten_1 (Flatten)	(None, 5376)	0
dense_6 (Dense)	(None, 64)	344128
dense_7 (Dense)	(None, 1)	65
Total params: 354,721		
Trainable params: 354,721		
Non-trainable params: 0		

```
In [74]: # Define the optimizer and loss:
model.compile(loss='mse',optimizer='adam')
# Define the callback to save the best model during the training
mc = ModelCheckpoint('best_model.hdf5', monitor='val_loss', verbose=1,
                    save_best_only=True, mode='min')
# Train the model for 30 epochs with batch size of 32:
history=model.fit(x_tr, y_tr ,epochs=30, batch_size=32, validation_data=(x_val,y_val),
                callbacks=[mc])
```

```
Epoch 1/30
133/133 [=====] - ETA: 0s - loss: 0.0971
Epoch 1: val_loss improved from inf to 0.04090, saving model to best_model.hdf5
133/133 [=====] - 3s 19ms/step - loss: 0.0971 - val_loss: 0.0409
Epoch 2/30
131/133 [=====>.] - ETA: 0s - loss: 0.0238
Epoch 2: val_loss improved from 0.04090 to 0.02447, saving model to best_model.hdf5
133/133 [=====] - 3s 21ms/step - loss: 0.0238 - val_loss: 0.0245
Epoch 3/30
131/133 [=====>.] - ETA: 0s - loss: 0.0172
Epoch 3: val_loss improved from 0.02447 to 0.01884, saving model to best_model.hdf5
133/133 [=====] - 3s 20ms/step - loss: 0.0174 - val_loss: 0.0188
Epoch 4/30
130/133 [=====>.] - ETA: 0s - loss: 0.0167
Epoch 4: val_loss improved from 0.01884 to 0.01482, saving model to best_model.hdf5
133/133 [=====] - 3s 20ms/step - loss: 0.0165 - val_loss: 0.0148
Epoch 5/30
131/133 [=====>.] - ETA: 0s - loss: 0.0127
```

Epoch 5: val_loss did not improve from 0.01482
133/133 [=====] - 3s 19ms/step - loss: 0.0126 - val_loss: 0.0164
Epoch 6/30
132/133 [=====>.] - ETA: 0s - loss: 0.0118
Epoch 6: val_loss improved from 0.01482 to 0.01394, saving model to best_model.hdf5
133/133 [=====] - 3s 20ms/step - loss: 0.0119 - val_loss: 0.0139
Epoch 7/30
131/133 [=====>.] - ETA: 0s - loss: 0.0112
Epoch 7: val_loss did not improve from 0.01394
133/133 [=====] - 3s 20ms/step - loss: 0.0112 - val_loss: 0.0155
Epoch 8/30
133/133 [=====] - ETA: 0s - loss: 0.0105
Epoch 8: val_loss improved from 0.01394 to 0.01391, saving model to best_model.hdf5
133/133 [=====] - 3s 20ms/step - loss: 0.0105 - val_loss: 0.0139
Epoch 9/30
133/133 [=====] - ETA: 0s - loss: 0.0105
Epoch 9: val_loss did not improve from 0.01391
133/133 [=====] - 3s 21ms/step - loss: 0.0105 - val_loss: 0.0140
Epoch 10/30
132/133 [=====>.] - ETA: 0s - loss: 0.0087
Epoch 10: val_loss improved from 0.01391 to 0.01337, saving model to best_model.hdf5
133/133 [=====] - 3s 21ms/step - loss: 0.0087 - val_loss: 0.0134
Epoch 11/30
133/133 [=====] - ETA: 0s - loss: 0.0084
Epoch 11: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0084 - val_loss: 0.0147
Epoch 12/30
131/133 [=====>.] - ETA: 0s - loss: 0.0090
Epoch 12: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0090 - val_loss: 0.0137
Epoch 13/30
132/133 [=====>.] - ETA: 0s - loss: 0.0078
Epoch 13: val_loss did not improve from 0.01337
133/133 [=====] - 3s 20ms/step - loss: 0.0078 - val_loss: 0.0137
Epoch 14/30
131/133 [=====>.] - ETA: 0s - loss: 0.0064
Epoch 14: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0064 - val_loss: 0.0146
Epoch 15/30
132/133 [=====>.] - ETA: 0s - loss: 0.0059
Epoch 15: val_loss did not improve from 0.01337
133/133 [=====] - 2s 19ms/step - loss: 0.0059 - val_loss: 0.0153
Epoch 16/30
131/133 [=====>.] - ETA: 0s - loss: 0.0059
Epoch 16: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0058 - val_loss: 0.0167
Epoch 17/30
131/133 [=====>.] - ETA: 0s - loss: 0.0049
Epoch 17: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0049 - val_loss: 0.0152
Epoch 18/30
131/133 [=====>.] - ETA: 0s - loss: 0.0043
Epoch 18: val_loss did not improve from 0.01337
133/133 [=====] - 2s 19ms/step - loss: 0.0043 - val_loss: 0.0165
Epoch 19/30
133/133 [=====] - ETA: 0s - loss: 0.0041
Epoch 19: val_loss did not improve from 0.01337
133/133 [=====] - 3s 20ms/step - loss: 0.0041 - val_loss: 0.0164
Epoch 20/30
133/133 [=====] - ETA: 0s - loss: 0.0038
Epoch 20: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0038 - val_loss: 0.0155
Epoch 21/30
132/133 [=====>.] - ETA: 0s - loss: 0.0033
Epoch 21: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0033 - val_loss: 0.0163
Epoch 22/30
133/133 [=====] - ETA: 0s - loss: 0.0028
Epoch 22: val_loss did not improve from 0.01337
133/133 [=====] - 2s 19ms/step - loss: 0.0028 - val_loss: 0.0164
Epoch 23/30
132/133 [=====>.] - ETA: 0s - loss: 0.0025
Epoch 23: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0025 - val_loss: 0.0165


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Epoch 24/30
133/133 [=====] - ETA: 0s - loss: 0.0035
Epoch 24: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0035 - val_loss: 0.0155
Epoch 25/30
131/133 [=====>.] - ETA: 0s - loss: 0.0030
Epoch 25: val_loss did not improve from 0.01337
133/133 [=====] - 2s 19ms/step - loss: 0.0030 - val_loss: 0.0158
Epoch 26/30
131/133 [=====>.] - ETA: 0s - loss: 0.0019
Epoch 26: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0019 - val_loss: 0.0174
Epoch 27/30
132/133 [=====>.] - ETA: 0s - loss: 0.0018
Epoch 27: val_loss did not improve from 0.01337
133/133 [=====] - 3s 20ms/step - loss: 0.0018 - val_loss: 0.0156
Epoch 28/30
131/133 [=====>.] - ETA: 0s - loss: 0.0016
Epoch 28: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0016 - val_loss: 0.0179
Epoch 29/30
132/133 [=====>.] - ETA: 0s - loss: 0.0013
Epoch 29: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0013 - val_loss: 0.0176
Epoch 30/30
131/133 [=====>.] - ETA: 0s - loss: 0.0015
Epoch 30: val_loss did not improve from 0.01337
133/133 [=====] - 3s 19ms/step - loss: 0.0015 - val_loss: 0.0154

```

```
In [75]: model.load_weights('best_model.hdf5')
```

```
In [76]: mse = model.evaluate(x_val,y_val)
print("Mean Square Error:",mse)
```

```

15/15 [=====] - 0s 5ms/step - loss: 0.0134
Mean Square Error: 0.013367628678679466

```

```
In [77]: #build a simple model
def compute_moving_average(data):
    pred=[]
    for i in data:
        avg=np.sum(i)/len(i)
        pred.append(avg)
    return np.array(pred)
x_resaped = x_val.reshape(-1,168)
y_pred = compute_moving_average(x_resaped)
mse = np.sum ( (y_val - y_pred) **2 ) / (len(y_val))
print("Mean Square Error:",mse)
```

```
Mean Square Error: 0.5546025834434455
```

```
In [78]: def forecast(x_val, no_of_pred, ind):
    predictions=[]
    #intialize the array with previous weeks data
    temp=x_val[ind]
    for i in range(no_of_pred):
        #predict for the next hour
        pred=model.predict(temp.reshape(1,-1,1))[0][0]

        #append the prediction as the last element of array
        temp = np.insert(temp,len(temp),pred)
        predictions.append(pred)
        #ignore the first element of array
        temp = temp[1:]
    return predictions
```

```
In [79]: no_of_pred =24
ind=72
y_pred= forecast(x_val,no_of_pred,ind)
```

```
y_true = y_val[ind:ind+(no_of_pred)]
```

```
1/1 [=====] - 0s 181ms/step
1/1 [=====] - 0s 43ms/step
1/1 [=====] - 0s 38ms/step
1/1 [=====] - 0s 46ms/step
1/1 [=====] - 0s 49ms/step
1/1 [=====] - 0s 40ms/step
1/1 [=====] - 0s 42ms/step
1/1 [=====] - 0s 43ms/step
1/1 [=====] - 0s 33ms/step
1/1 [=====] - 0s 43ms/step
1/1 [=====] - 0s 37ms/step
1/1 [=====] - 0s 43ms/step
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 34ms/step
1/1 [=====] - 0s 42ms/step
1/1 [=====] - 0s 30ms/step
1/1 [=====] - 0s 33ms/step
1/1 [=====] - 0s 42ms/step
1/1 [=====] - 0s 46ms/step
1/1 [=====] - 0s 42ms/step
1/1 [=====] - 0s 47ms/step
1/1 [=====] - 0s 32ms/step
1/1 [=====] - 0s 40ms/step
```

```
In [80]: y_true = np.array(y_true)
y_true = y_true.reshape(-1, 1)

y_pred = np.array(y_pred)
y_pred = y_pred.reshape(-1, 1)
y_true= y_scaler.inverse_transform(y_true)
y_pred= y_scaler.inverse_transform(y_pred)
```

```
In [81]: def plot(y_true,y_pred):
ar = np.arange(len(y_true))
plt.figure(figsize=(22,10))
plt.plot(ar, y_true,'r')
plt.plot(ar, y_pred,'y')
plt.show()
```

```
In [82]: plot(y_true,y_pred)
```

