CIND820 Initial Coding

December 4, 2020

```
[1]: import os
     import warnings
     warnings.filterwarnings('ignore')
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     plt.style.use('fivethirtyeight')
     from pylab import rcParams
     rcParams['figure.figsize'] = 10, 6
     from datetime import datetime
     from statsmodels.tsa.stattools import adfuller
     from statsmodels.tsa.seasonal import seasonal_decompose
     from statsmodels.tsa.arima_model import ARIMA
     from sklearn.metrics import mean squared error, mean_absolute_error
     import math
     from statsmodels.tsa.stattools import acf, pacf
```

[2]: pip install pmdarima

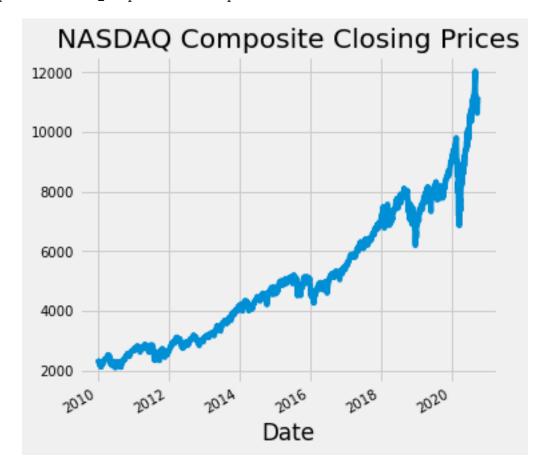
```
Requirement already satisfied: pmdarima in /opt/conda/lib/python3.7/site-
packages (1.8.0)
Requirement already satisfied: numpy>=1.17.3 in /opt/conda/lib/python3.7/site-
packages (from pmdarima) (1.18.4)
Requirement already satisfied: scipy>=1.3.2 in /opt/conda/lib/python3.7/site-
packages (from pmdarima) (1.4.1)
Requirement already satisfied: Cython<0.29.18,>=0.29 in
/opt/conda/lib/python3.7/site-packages (from pmdarima) (0.29.17)
Requirement already satisfied: setuptools!=50.0.0,>=38.6.0 in
/opt/conda/lib/python3.7/site-packages (from pmdarima) (46.1.3.post20200325)
Requirement already satisfied: pandas>=0.19 in /opt/conda/lib/python3.7/site-
packages (from pmdarima) (1.0.3)
Requirement already satisfied: joblib>=0.11 in /opt/conda/lib/python3.7/site-
packages (from pmdarima) (0.15.1)
Requirement already satisfied: urllib3 in /opt/conda/lib/python3.7/site-packages
(from pmdarima) (1.25.9)
Requirement already satisfied: scikit-learn>=0.22 in
/opt/conda/lib/python3.7/site-packages (from pmdarima) (0.22.2.post1)
Requirement already satisfied: statsmodels!=0.12.0,>=0.11 in
```

```
Requirement already satisfied: pytz>=2017.2 in /opt/conda/lib/python3.7/site-
      packages (from pandas>=0.19->pmdarima) (2020.1)
      Requirement already satisfied: python-dateutil>=2.6.1 in
      /opt/conda/lib/python3.7/site-packages (from pandas>=0.19->pmdarima) (2.8.1)
      Requirement already satisfied: patsy>=0.5 in /opt/conda/lib/python3.7/site-
      packages (from statsmodels!=0.12.0,>=0.11->pmdarima) (0.5.1)
      Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.7/site-
      packages (from python-dateutil>=2.6.1->pandas>=0.19->pmdarima) (1.14.0)
      Note: you may need to restart the kernel to use updated packages.
 [3]: from pmdarima.arima import auto_arima
[132]: #import NASDAQ data
      df=pd.read csv("IXIC v1.csv", sep=",")
[133]: #understand data format and clean up data
      from datetime import datetime
      con=df['Date']
      df['Date'] = pd.to_datetime(df['Date'])
      df.set_index('Date', inplace=True)
       #check datatype of index
      df.index
[133]: DatetimeIndex(['2010-01-04', '2010-01-05', '2010-01-06', '2010-01-07',
                      '2010-01-08', '2010-01-11', '2010-01-12', '2010-01-13',
                      '2010-01-14', '2010-01-15',
                      '2020-09-17', '2020-09-18', '2020-09-21', '2020-09-22',
                      '2020-09-23', '2020-09-24', '2020-09-25', '2020-09-28',
                      '2020-09-29', '2020-09-30'],
                     dtype='datetime64[ns]', name='Date', length=2705, freq=None)
[134]: df['year'] = df.index.year
      df['month'] = df.index.month
      df['day'] = df.index.day
[135]: df.head()
[135]:
                        Close year month day
      Date
      2010-01-04 2308.419922 2010
                                          1
      2010-01-05 2308.709961 2010
                                          1
                                               5
      2010-01-06 2301.090088 2010
                                          1
                                               6
                                              7
      2010-01-07 2300.050049 2010
                                          1
      2010-01-08 2317.169922 2010
```

/opt/conda/lib/python3.7/site-packages (from pmdarima) (0.11.1)

```
[136]: #plot NASDAQ trend
temp=df.groupby(['Date'])['Close'].mean()
temp.plot(figsize=(5,5), title= 'NASDAQ Composite Closing Prices', fontsize=10)
```

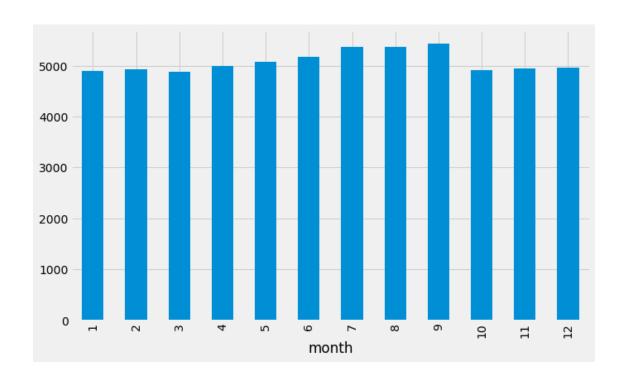
[136]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde86daad50>



[137]: df.groupby('month')['Close'].mean().plot.bar()
#on average, september has the highest average price compares to the other_

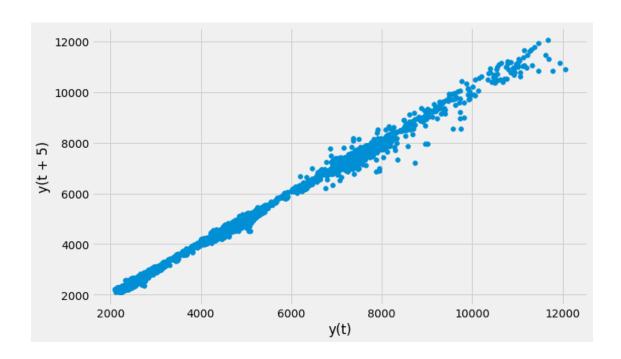
--months.

[137]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde86d9e190>



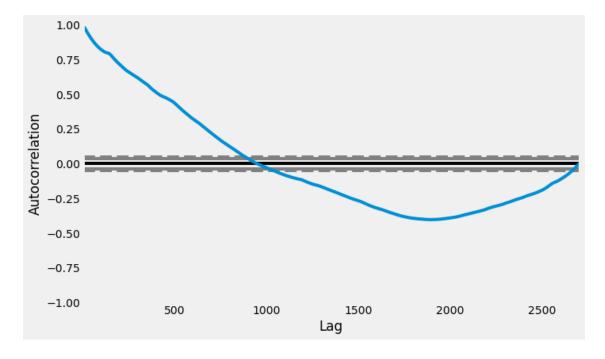
[138]: #lag plot from pandas.plotting import lag_plot lag_plot(df['Close'],lag=5) #Graph shows a linear pattern. Implies data points are non random and suggests → that an autoregressive model might be appropriate.

[138]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde86d163d0>



[139]: from pandas.plotting import autocorrelation_plot autocorrelation_plot(df['Close']) #there is high level of correlation

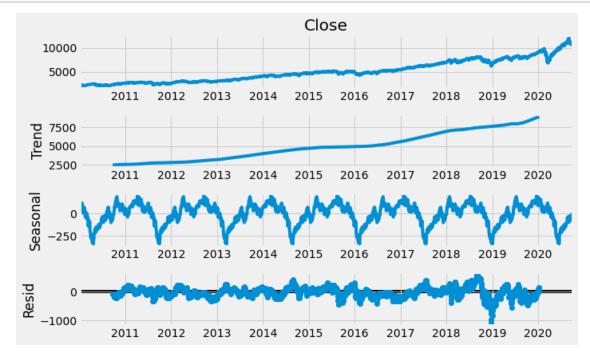
[139]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde86c8a410>



```
[140]: #data is not stationary based on high p value
from statsmodels.tsa.stattools import adfuller
result = adfuller(df.Close.dropna())
print(f"ADF Statstic: {result[0]}")
print(f"p-value:{result[1]}")
```

ADF Statstic: 1.4430465972942679 p-value:0.9973011850493003

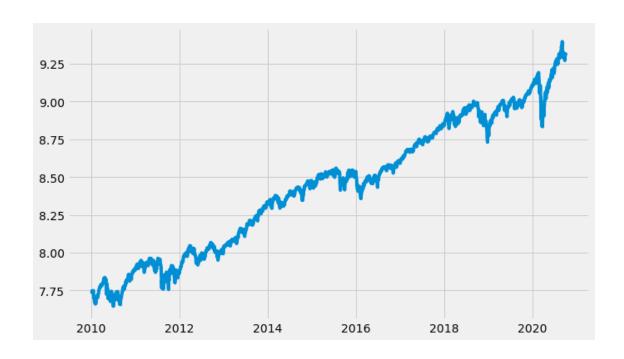
```
[141]: #decompose data
import statsmodels.api as sm
res = sm.tsa.seasonal_decompose(df['Close'],model= 'addictive',period = 365)
resplot = res.plot()
#data shows upward trend and presents seasonlity
```



```
[142]: def plot_df(df,x,y,title= "", xlabel = "Date", ylabel='Value',dpi=50):
    plt.plot(x,y)
    plt.show()
```

```
[143]: #apply log transformation to stablize data plt.plot(df.apply(np.log)['Close'])
```

[143]: [<matplotlib.lines.Line2D at 0x7fdeb01b89d0>]

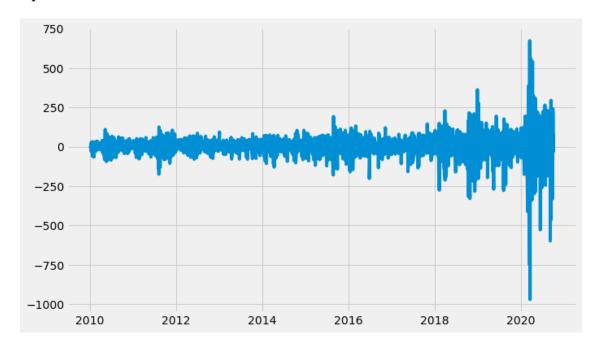


[144]: #To covert data into stationary dataset, first differencing has to be applied.

With first differencing, empty field needs to be filled as 0.

plt.plot(df['Close'].diff(1).fillna(0))

[144]: [<matplotlib.lines.Line2D at 0x7fde84a13c90>]

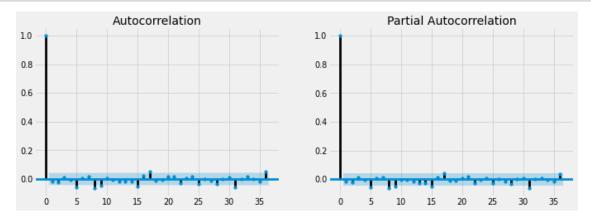


```
[145]: #confirm stationarity
      from statsmodels.tsa.stattools import adfuller
      result = adfuller(np.log(df['Close']).diff(1).fillna(0))
      print(f"ADF Statstic: {result[0]}")
      print(f"p-value:{result[1]}")
      ADF Statstic: -11.770291597674845
      p-value:1.0940406662618215e-21
[146]: df_st= df.diff(1).fillna(0)
[147]: import time
      t = time.process_time()
[148]: #understand the structure of the stationary dataset
      df_st.head()
[148]:
                      Close year month day
      Date
      2010-01-04 0.000000
                              0.0
                                     0.0 0.0
      2010-01-05 0.290039
                              0.0
                                     0.0 1.0
      2010-01-06 -7.619873
                              0.0
                                     0.0 1.0
      2010-01-07 -1.040039
                              0.0
                                     0.0 1.0
      2010-01-08 17.119873
                              0.0
                                     0.0 1.0
[149]: | #With transformation to maintain stationarity, we need to be model back to the
       →original dataset in order to predict stock price.
      df_revert=df_st.copy()
      df_revert=df_revert.cumsum()
[150]: df_revert.iloc[0,:]=df.iloc[0,:]
      df_revert = df_revert.cumsum()
[151]: #define data
      class TimeSeriesData():
          def init (self, df):
              self.data = df
              self.stationary = self.stationarize(df)
               self.revert = self.revert(self.stationary, self.data)
          def revert(self, st, org):
              x = st.copy()
              x.iloc[0,:] = org.iloc[0,:]
              return x.cumsum()
          def stationarize(self, data):
```

return data.diff(1).fillna(0)

```
[152]: #split dataset
x_train = TimeSeriesData(df[:int((len(df)*0.8))])
x_test = TimeSeriesData(df[int((len(df)*0.8)):])
```

[153]: #plot ACF and PACF for the stationary dataset
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
fig, axes = plt.subplots(1,2,figsize=(15,5), dpi= 50)
plot_acf(x_train.stationary['Close'].values.tolist(), lags=36, ax=axes[0]);
plot_pacf(x_train.stationary['Close'].values.tolist(), lags=36, ax=axes[1]);



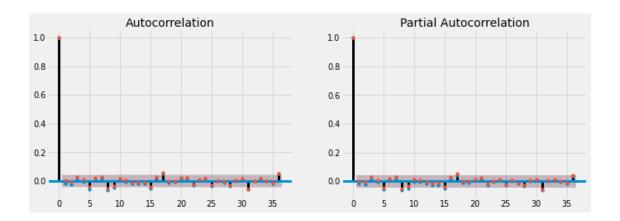
Performing stepwise search to minimize aic

: AIC=22406.770, Time=0.74 sec ARIMA(2,1,2)(0,0,0)[0] intercept : AIC=22400.729, Time=0.05 sec ARIMA(0,1,0)(0,0,0)[0] intercept ARIMA(1,1,0)(0,0,0)[0] intercept : AIC=22402.090, Time=0.09 sec ARIMA(0,1,1)(0,0,0)[0] intercept : AIC=22402.065, Time=0.13 sec : AIC=22406.532, Time=0.04 sec ARIMA(0,1,0)(0,0,0)[0]ARIMA(1,1,1)(0,0,0)[0] intercept : AIC=22393.839, Time=1.27 sec ARIMA(2,1,1)(0,0,0)[0] intercept : AIC=22404.619, Time=0.61 sec ARIMA(1,1,2)(0,0,0)[0] intercept : AIC=22404.702, Time=0.83 sec ARIMA(0,1,2)(0,0,0)[0] intercept : AIC=22402.822, Time=0.43 sec ARIMA(2,1,0)(0,0,0)[0] intercept : AIC=22402.801, Time=0.16 sec ARIMA(1,1,1)(0,0,0)[0]: AIC=22406.446, Time=0.28 sec

Best model: ARIMA(1,1,1)(0,0,0)[0] intercept

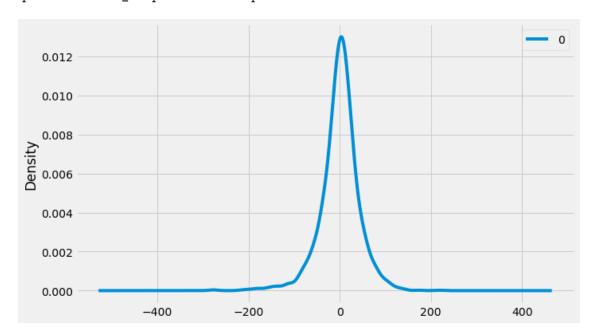
Total fit time: 4.629 seconds

```
[154]: ARIMA(maxiter=50, method='lbfgs', order=(1, 1, 1), out_of_sample_size=0,
          scoring='mse', scoring_args={}, seasonal_order=(0, 0, 0, 0),
          start_params=None, suppress_warnings=True, trend=None,
          with_intercept=True)
[155]: model_arima = ARIMA(x_train.data['Close'].values, order=(1,1,1))
[156]: result_arima = model_arima.fit(disp=-1)
[157]: print(result_arima.summary())
                           ARIMA Model Results
     ______
    Dep. Variable:
                               D.y
                                    No. Observations:
                                                               2163
    Model:
                      ARIMA(1, 1, 1) Log Likelihood
                                                         -11192.925
    Method:
                            css-mle S.D. of innovations
                                                             42.767
    Date:
                    Fri, 04 Dec 2020 AIC
                                                           22393.851
                           19:12:02 BIC
    Time:
                                                           22416.568
    Sample:
                              1 HQIC
                                                           22402.159
                                           P>|z|
                                                    [0.025
                 coef
                        std err
                                                             0.975
     ______
                2.5691
                          0.570
                                  4.506
                                           0.000
                                                    1.452
                                                              3.687
                          0.027 34.756
                                          0.000
    ar.L1.D.y
               0.9330
                                                    0.880
                                                             0.986
    ma.L1.D.y
              -0.9587
                        0.021
                                -45.615
                                          0.000
                                                   -1.000
                                                             -0.917
                                 Roots
     ______
                             Imaginary
                                             Modulus
    AR.1
                1.0718
                              +0.0000j
                                             1.0718
                                                             0.0000
                              +0.0000j
                                              1.0431
    MA.1
                 1.0431
                                                             0.0000
[158]: #understand residual
     residuals = pd.DataFrame(result arima.resid)
[159]: plot_acf(residuals, lags=36, ax=axes[0])
     plot_pacf(residuals, lags=36, ax=axes[1])
[159]:
```

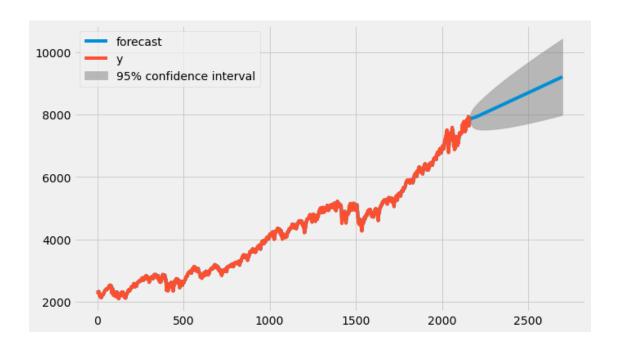


[160]: residuals.plot(kind='kde')

[160]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde848bcd90>



[161]: result_arima.plot_predict(1,2700);



```
[162]: prediction = result_arima.predict(len(df)-200,len(df)-1)
[163]: from sklearn.metrics import mean_absolute_error
       from sklearn.metrics import mean_squared_error
       from sklearn.metrics import r2_score
       mae = mean_absolute_error(x_test.data['Close'], np.append(x_train.data.iloc[-1,:
       → ]['Close'], prediction).cumsum()[1:])
       mse = mean_squared_error(x_test.data['Close'], np.append(x_train.data.iloc[-1,:
       →]['Close'], prediction).cumsum()[1:])
       rmse = np.sqrt(mse)
       print("Results of sklearn.metrics:")
       print("MAE:",mae)
       print("MSE:", mse)
       print("RMSE:", rmse)
      Results of sklearn.metrics:
      MAE: 690.3840484817343
      MSE: 791528.2196244474
      RMSE: 889.678717079625
[164]: def smape(a, f):
           return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)
       smape(x_test.data['Close'], np.append(x_train.data.iloc[-1,:]['Close'],__
        →prediction).cumsum()[1:])
```

```
[164]: 7.997544256633192
[165]: elapsed_time = time.process_time() - t
       print(elapsed_time)
      18.974526218007668
  []:
  []:
  []:
  []:
  []:
  []:
[37]:
       #same analysis for TSX price
[166]: #upload TSX price
       df=pd.read_csv("GSPTSE_v1.csv", sep=",")
[167]: #understand the data and covert date format
       from datetime import datetime
       con=df['Date']
       df['Date'] = pd.to_datetime(df['Date'])
       df.set index('Date', inplace=True)
       #check datatype of index
       df.index
[167]: DatetimeIndex(['2010-01-04', '2010-01-05', '2010-01-06', '2010-01-07',
                      '2010-01-08', '2010-01-11', '2010-01-12', '2010-01-13',
                      '2010-01-14', '2010-01-15',
                      '2020-09-17', '2020-09-18', '2020-09-21', '2020-09-22',
                      '2020-09-23', '2020-09-24', '2020-09-25', '2020-09-28',
                      '2020-09-29', '2020-09-30'],
                     dtype='datetime64[ns]', name='Date', length=2697, freq=None)
[168]: df['year'] = df.index.year
       df['month'] = df.index.month
       df['day'] = df.index.day
[169]: df.head()
```

```
Close year month day
      Date
      2010-01-04 11866.90039
                              2010
                                             4
      2010-01-05 11888.09961 2010
                                             5
      2010-01-06 11944.50000 2010
                                             6
      2010-01-07 11887.50000 2010
                                             7
      2010-01-08 11953.79981 2010
[170]: #plot TSX trend
      temp=df.groupby(['Date'])['Close'].mean()
      temp.plot(figsize=(5,5), title= 'TSX Composite Closing Prices', fontsize=10)
```

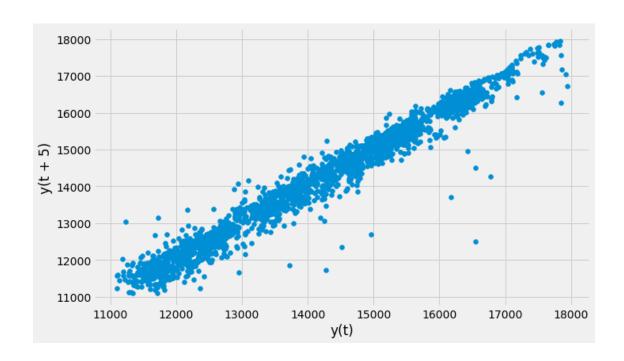
[170]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde86da99d0>

[169]:



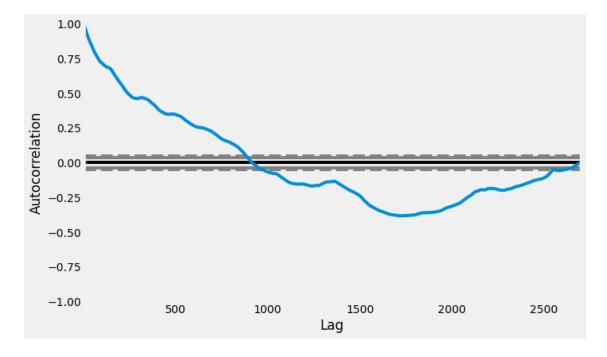
```
[171]: #lag plot
       from pandas.plotting import lag_plot
       lag_plot(df['Close'],lag=5)
       #a linear plot also indicates non random dataset
```

[171]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde848dc550>



[172]: from pandas.plotting import autocorrelation_plot autocorrelation_plot(df['Close']) #autocorrelation suggests arima might be a good model

[172]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde8459c410>



```
[173]: #original data shows as not stationary due to high p level
    from statsmodels.tsa.stattools import adfuller
    result = adfuller(df.Close.dropna())
    print(f"ADF Statstic: {result[0]}")
    print(f"p-value:{result[1]}")
```

ADF Statstic: -2.1272017170757755 p-value:0.23371823999228158

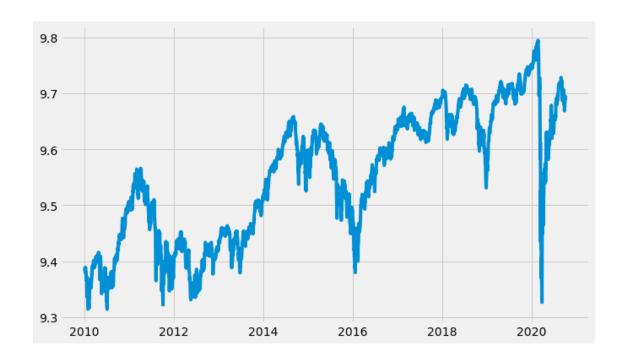
```
[174]: #decompose data
import statsmodels.api as sm
res = sm.tsa.seasonal_decompose(df['Close'],model= 'addictive',period = 365)
resplot = res.plot()
#upward trend
```



```
[175]: def plot_df(df,x,y,title= "", xlabel = "Date", ylabel='Value',dpi=50):
    plt.plot(x,y)
    plt.show()
```

```
[176]: #apply log transformation to stablize data plt.plot(df.apply(np.log)['Close'])
```

[176]: [<matplotlib.lines.Line2D at 0x7fde843b21d0>]

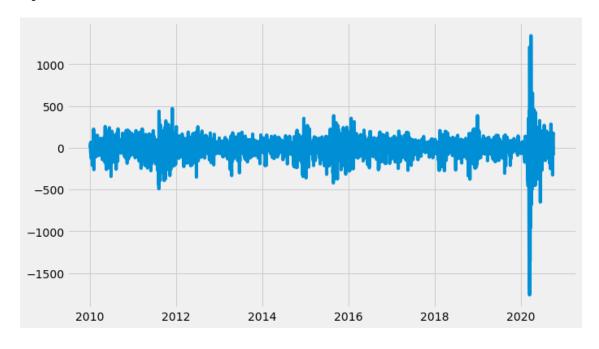


[177]: #To covert data into stationary dataset, first differencing has to be applied.

With first differencing, empty field needs to be filled as 0.

plt.plot(df['Close'].diff(1).fillna(0))

[177]: [<matplotlib.lines.Line2D at 0x7fde84391e50>]

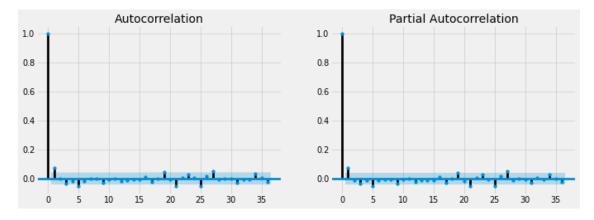


```
[178]: #confirm stationarity
      from statsmodels.tsa.stattools import adfuller
      result = adfuller(np.log(df['Close']).diff(1).fillna(0))
      print(f"ADF Statstic: {result[0]}")
      print(f"p-value:{result[1]}")
      ADF Statstic: -10.92108896454532
      p-value:1.037610106877016e-19
[179]: import time
      t = time.process_time()
[180]: df_st= df.diff(1).fillna(0)
[181]: #understand the structure of the stationary dataset
      df_st.head()
[181]:
                     Close year month day
      Date
      2010-01-04 0.00000 0.0
                                    0.0 0.0
      2010-01-05 21.19922 0.0
                                    0.0 1.0
      2010-01-06 56.40039 0.0
                                    0.0 1.0
      2010-01-07 -57.00000 0.0
                                    0.0 1.0
      2010-01-08 66.29981 0.0
                                    0.0 1.0
[182]: #With transformation to maintain stationarity, we need to be model back to the
       →original dataset in order to predict stock price.
      df_revert=df_st.copy()
      df_revert=df_revert.cumsum()
[183]: df_revert.iloc[0,:]=df.iloc[0,:]
      df_revert = df_revert.cumsum()
[184]: #define data
      class TimeSeriesData():
          def init (self, df):
              self.data = df
              self.stationary = self.stationarize(df)
              self.revert = self.revert(self.stationary, self.data)
          def revert(self, st, org):
              x = st.copy()
              x.iloc[0,:] = org.iloc[0,:]
              return x.cumsum()
          def stationarize(self, data):
```

return data.diff(1).fillna(0)

```
[185]: #split dataset
x_train = TimeSeriesData(df[:int((len(df)*0.85))])
x_test = TimeSeriesData(df[int((len(df)*0.85)):])
```

[186]: #plot ACF and PACF for the stationary dataset
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
fig, axes = plt.subplots(1,2,figsize=(15,5), dpi= 50)
plot_acf(x_train.stationary['Close'].values.tolist(), lags=36,ax=axes[0]);
plot_pacf(x_train.stationary['Close'].values.tolist(), lags=36,ax=axes[1]);



Performing stepwise search to minimize aic

```
: AIC=27707.940, Time=1.33 sec
ARIMA(2,1,2)(0,0,0)[0] intercept
                                   : AIC=27722.152, Time=0.05 sec
ARIMA(0,1,0)(0,0,0)[0] intercept
ARIMA(1,1,0)(0,0,0)[0] intercept
                                   : AIC=27711.070, Time=0.10 sec
ARIMA(0,1,1)(0,0,0)[0] intercept
                                   : AIC=27711.001, Time=0.13 sec
                                   : AIC=27720.872, Time=0.03 sec
ARIMA(0,1,0)(0,0,0)[0]
ARIMA(1,1,2)(0,0,0)[0] intercept
                                  : AIC=27714.994, Time=0.23 sec
ARIMA(2,1,1)(0,0,0)[0] intercept
                                   : AIC=27707.115, Time=1.35 sec
ARIMA(1,1,1)(0,0,0)[0] intercept
                                   : AIC=27712.993, Time=0.13 sec
ARIMA(2,1,0)(0,0,0)[0] intercept
                                   : AIC=27712.917, Time=0.16 sec
ARIMA(3,1,1)(0,0,0)[0] intercept
                                   : AIC=27707.866, Time=1.59 sec
ARIMA(3,1,0)(0,0,0)[0] intercept
                                   : AIC=27712.282, Time=0.22 sec
ARIMA(3,1,2)(0,0,0)[0] intercept
                                   : AIC=27708.220, Time=2.16 sec
ARIMA(2,1,1)(0,0,0)[0]
                                   : AIC=27706.090, Time=0.43 sec
ARIMA(1,1,1)(0,0,0)[0]
                                   : AIC=27711.613, Time=0.19 sec
```

```
ARIMA(2,1,0)(0,0,0)[0] : AIC=27711.544, Time=0.06 sec

ARIMA(3,1,1)(0,0,0)[0] : AIC=27706.835, Time=0.42 sec

ARIMA(2,1,2)(0,0,0)[0] : AIC=27706.903, Time=0.47 sec

ARIMA(1,1,0)(0,0,0)[0] : AIC=27709.688, Time=0.05 sec

ARIMA(1,1,2)(0,0,0)[0] : AIC=27713.616, Time=0.11 sec

ARIMA(3,1,0)(0,0,0)[0] : AIC=27710.951, Time=0.09 sec

ARIMA(3,1,2)(0,0,0)[0] : AIC=27707.194, Time=0.73 sec
```

Best model: ARIMA(2,1,1)(0,0,0)[0] Total fit time: 10.038 seconds

[187]: ARIMA(maxiter=50, method='lbfgs', order=(2, 1, 1), out_of_sample_size=0, scoring='mse', scoring_args={}, seasonal_order=(0, 0, 0, 0), start_params=None, suppress_warnings=True, trend=None, with_intercept=False)

```
[188]: model_arima = ARIMA(x_train.data['Close'].values, order=(2,1,1))
```

[189]: result_arima = model_arima.fit(disp=-1)

[190]: print(result_arima.summary())

ARIMA Model Results

Dep. Variable:	D.y	No. Observations:	2291
Model:	ARIMA(2, 1, 1)	Log Likelihood	-13848.621
Method:	css-mle	S.D. of innovations	102.089
Date:	Fri, 04 Dec 2020	AIC	27707.242
Time:	19:12:38	BIC	27735.926
Sample:	1	HQIC	27717.702

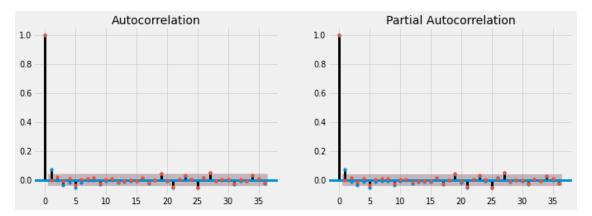
=========	:========	========	========	:========	=========	=======
	coef	std err	z	P> z	[0.025	0.975]
const	1.7918	1.819	0.985	0.325	-1.773	5.357
ar.L1.D.y	0.9619	0.058	16.567	0.000	0.848	1.076
ar.L2.D.y	-0.0939	0.021	-4.491	0.000	-0.135	-0.053
ma.L1.D.y	-0.8876	0.055	-16.179	0.000	-0.995	-0.780
			Roots			

	Real	Imaginary	Modulus	Frequency
AR.1	1.1741	+0.0000j	1.1741	0.0000
AR.2	9.0727	+0.0000j	9.0727	0.0000
MA.1	1.1267	+0.0000j	1.1267	0.0000

```
[191]: #understand residual
residuals = pd.DataFrame(result_arima.resid)
```

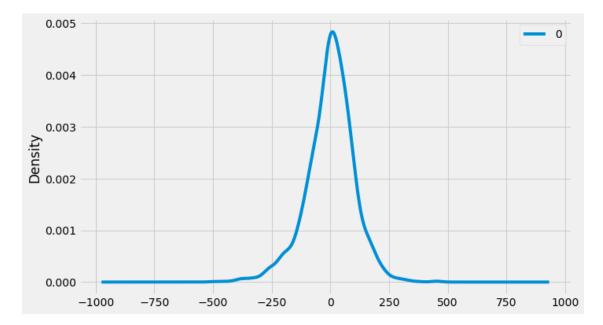
```
[192]: plot_acf(residuals, lags=36, ax=axes[0]) plot_pacf(residuals, lags=36, ax=axes[1])
```



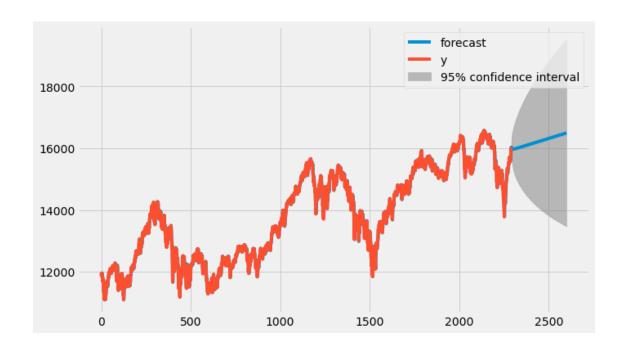


```
[193]: residuals.plot(kind='kde')
```

[193]: <matplotlib.axes._subplots.AxesSubplot at 0x7fde8422d150>



```
[194]: result_arima.plot_predict(1,2600);
```



```
[195]: prediction = result_arima.predict(len(df)-208,len(df)-1)
[196]: from sklearn.metrics import mean_absolute_error
       from sklearn.metrics import mean_squared_error
       from sklearn.metrics import r2_score
       mae = mean_absolute_error(x_test.data['Close'], np.append(x_train.data.iloc[-1,:
       →]['Close'], prediction).cumsum()[1:])
       mse = mean_squared_error(x_test.data['Close'], np.append(x_train.data.iloc[-1,:
       →]['Close'], prediction).cumsum()[1:])
       rmse = np.sqrt(mse)
       print("Results of sklearn.metrics:")
       print("MAE:",mae)
       print("MSE:", mse)
       print("RMSE:", rmse)
      Results of sklearn.metrics:
      MAE: 767.5063223410224
      MSE: 1313293.7357536359
      RMSE: 1145.9902860642562
[197]: def smape(a, f):
           return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)
       smape(x_test.data['Close'], np.append(x_train.data.iloc[-1,:]['Close'],__
        →prediction).cumsum()[1:])
```

```
[197]: 4.863946788376522
[198]: elapsed_time = time.process_time() - t
       print(elapsed_time)
      34.980218798998976
 []:
 []:
       #using LSTM to predict stock price
[70]:
[71]: pip install tensorflow
      Requirement already satisfied: tensorflow in /opt/conda/lib/python3.7/site-
      packages (2.3.1)
      Requirement already satisfied: gast==0.3.3 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (0.3.3)
      Requirement already satisfied: numpy<1.19.0,>=1.16.0 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.18.4)
      Requirement already satisfied: opt-einsum>=2.3.2 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (3.3.0)
      Requirement already satisfied: tensorflow-estimator<2.4.0,>=2.3.0 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (2.3.0)
      Requirement already satisfied: termcolor>=1.1.0 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.1.0)
      Requirement already satisfied: wrapt>=1.11.1 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (1.12.1)
      Requirement already satisfied: wheel>=0.26 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (0.34.2)
      Requirement already satisfied: astunparse==1.6.3 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.6.3)
      Requirement already satisfied: absl-py>=0.7.0 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (0.11.0)
      Requirement already satisfied: h5py<2.11.0,>=2.10.0 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (2.10.0)
      Requirement already satisfied: keras-preprocessing<1.2,>=1.1.1 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.1.2)
      Requirement already satisfied: protobuf>=3.9.2 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (3.11.4)
      Requirement already satisfied: google-pasta>=0.1.8 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (0.2.0)
      Requirement already satisfied: tensorboard<3,>=2.3.0 in
      /opt/conda/lib/python3.7/site-packages (from tensorflow) (2.4.0)
      Requirement already satisfied: grpcio>=1.8.6 in /opt/conda/lib/python3.7/site-
      packages (from tensorflow) (1.33.2)
      Requirement already satisfied: six>=1.12.0 in /opt/conda/lib/python3.7/site-
```

```
packages (from tensorflow) (1.14.0)
Requirement already satisfied: setuptools in /opt/conda/lib/python3.7/site-
packages (from protobuf>=3.9.2->tensorflow) (46.1.3.post20200325)
Requirement already satisfied: markdown>=2.6.8 in /opt/conda/lib/python3.7/site-
packages (from tensorboard<3,>=2.3.0->tensorflow) (3.3.3)
Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in
/opt/conda/lib/python3.7/site-packages (from tensorboard<3,>=2.3.0->tensorflow)
(1.7.0)
Requirement already satisfied: requests<3,>=2.21.0 in
/opt/conda/lib/python3.7/site-packages (from tensorboard<3,>=2.3.0->tensorflow)
(2.23.0)
Requirement already satisfied: google-auth<2,>=1.6.3 in
/opt/conda/lib/python3.7/site-packages (from tensorboard<3,>=2.3.0->tensorflow)
(1.16.1)
Requirement already satisfied: werkzeug>=0.11.15 in
/opt/conda/lib/python3.7/site-packages (from tensorboard<3,>=2.3.0->tensorflow)
(1.0.1)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in
/opt/conda/lib/python3.7/site-packages (from tensorboard<3,>=2.3.0->tensorflow)
(0.4.2)
Requirement already satisfied: importlib-metadata; python_version < "3.8" in
/opt/conda/lib/python3.7/site-packages (from
markdown>=2.6.8->tensorboard<3,>=2.3.0->tensorflow) (1.6.0)
Requirement already satisfied: certifi>=2017.4.17 in
/opt/conda/lib/python3.7/site-packages (from
requests<3,>=2.21.0->tensorboard<3,>=2.3.0->tensorflow) (2020.4.5.2)
Requirement already satisfied: chardet<4,>=3.0.2 in
/opt/conda/lib/python3.7/site-packages (from
requests<3,>=2.21.0->tensorboard<3,>=2.3.0->tensorflow) (3.0.4)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in
/opt/conda/lib/python3.7/site-packages (from
requests<3,>=2.21.0->tensorboard<3,>=2.3.0->tensorflow) (1.25.9)
Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.7/site-
packages (from requests<3,>=2.21.0->tensorboard<3,>=2.3.0->tensorflow) (2.9)
Requirement already satisfied: cachetools<5.0,>=2.0.0 in
/opt/conda/lib/python3.7/site-packages (from google-
auth<2,>=1.6.3->tensorboard<3,>=2.3.0->tensorflow) (4.1.0)
Requirement already satisfied: rsa<4.1,>=3.1.4 in /opt/conda/lib/python3.7/site-
packages (from google-auth<2,>=1.6.3->tensorboard<3,>=2.3.0->tensorflow) (4.0)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
/opt/conda/lib/python3.7/site-packages (from google-
auth<2,>=1.6.3->tensorboard<3,>=2.3.0->tensorflow) (0.2.8)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
/opt/conda/lib/python3.7/site-packages (from google-auth-
oauthlib<0.5,>=0.4.1->tensorboard<3,>=2.3.0->tensorflow) (1.3.0)
Requirement already satisfied: zipp>=0.5 in /opt/conda/lib/python3.7/site-
packages (from importlib-metadata; python_version <</pre>
"3.8"->markdown>=2.6.8->tensorboard<3,>=2.3.0->tensorflow) (3.1.0)
```

Requirement already satisfied: pyasn1>=0.1.3 in /opt/conda/lib/python3.7/site-packages (from rsa<4.1,>=3.1.4->google-auth<2,>=1.6.3->tensorboard<3,>=2.3.0->tensorflow) (0.4.8)

Requirement already satisfied: oauthlib>=3.0.0 in /opt/conda/lib/python3.7/site-packages (from requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard<3,>=2.3.0->tensorflow) (3.0.1)

Note: you may need to restart the kernel to use updated packages.

[72]: pip install keras

Requirement already satisfied: keras in /opt/conda/lib/python3.7/site-packages (2.4.3)

Requirement already satisfied: h5py in /opt/conda/lib/python3.7/site-packages (from keras) (2.10.0)

Requirement already satisfied: numpy>=1.9.1 in /opt/conda/lib/python3.7/site-packages (from keras) (1.18.4)

Requirement already satisfied: scipy>=0.14 in /opt/conda/lib/python3.7/site-packages (from keras) (1.4.1)

Requirement already satisfied: pyyaml in /opt/conda/lib/python3.7/site-packages (from keras) (5.3.1)

Requirement already satisfied: six in /opt/conda/lib/python3.7/site-packages

Note: you may need to restart the kernel to use updated packages.

[73]: pip install pandas-datareader

(from h5py->keras) (1.14.0)

Requirement already satisfied: pandas-datareader in /opt/conda/lib/python3.7/site-packages (0.9.0) Requirement already satisfied: lxml in /opt/conda/lib/python3.7/site-packages (from pandas-datareader) (4.5.1) Requirement already satisfied: requests>=2.19.0 in /opt/conda/lib/python3.7/site-packages (from pandas-datareader) (2.23.0) Requirement already satisfied: pandas>=0.23 in /opt/conda/lib/python3.7/sitepackages (from pandas-datareader) (1.0.3) Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /opt/conda/lib/python3.7/site-packages (from requests>=2.19.0->pandasdatareader) (1.25.9) Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.7/site-packages (from requests>=2.19.0->pandasdatareader) (2020.4.5.2) Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.7/sitepackages (from requests>=2.19.0->pandas-datareader) (2.9) Requirement already satisfied: chardet<4,>=3.0.2 in /opt/conda/lib/python3.7/site-packages (from requests>=2.19.0->pandasdatareader) (3.0.4) Requirement already satisfied: numpy>=1.13.3 in /opt/conda/lib/python3.7/sitepackages (from pandas>=0.23->pandas-datareader) (1.18.4) Requirement already satisfied: pytz>=2017.2 in /opt/conda/lib/python3.7/site-

```
packages (from pandas>=0.23->pandas-datareader) (2020.1)
     Requirement already satisfied: python-dateutil>=2.6.1 in
     /opt/conda/lib/python3.7/site-packages (from pandas>=0.23->pandas-datareader)
     Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.7/site-
     packages (from python-dateutil>=2.6.1->pandas>=0.23->pandas-datareader) (1.14.0)
     Note: you may need to restart the kernel to use updated packages.
[74]: pip install scikit-learn
     Requirement already satisfied: scikit-learn in /opt/conda/lib/python3.7/site-
     packages (0.22.2.post1)
     Requirement already satisfied: joblib>=0.11 in /opt/conda/lib/python3.7/site-
     packages (from scikit-learn) (0.15.1)
     Requirement already satisfied: scipy>=0.17.0 in /opt/conda/lib/python3.7/site-
     packages (from scikit-learn) (1.4.1)
     Requirement already satisfied: numpy>=1.11.0 in /opt/conda/lib/python3.7/site-
     packages (from scikit-learn) (1.18.4)
     Note: you may need to restart the kernel to use updated packages.
[75]: from pandas_datareader import data
      import datetime as dt
      from matplotlib import pyplot as plt
      from sklearn import model_selection
      from sklearn.metrics import confusion_matrix
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      import numpy as np
      import pandas as pd
      from sklearn.preprocessing import MinMaxScaler
      from keras.models import Sequential
      from keras.layers import Dense
      from keras.layers import LSTM
      from keras.layers import GRU
      from keras.layers import Dropout
[76]: #similar to before, upload NASDAQ data
      df=pd.read_csv("IXIC_v1.csv", sep=",")
      from datetime import datetime
      con=df['Date']
      df['Date'] = pd.to_datetime(df['Date'])
      df.set_index('Date', inplace=True)
```

```
[77]: df['Date'] = df.index
data2 = pd.DataFrame(columns = ['Date', 'Close'])
```

test = df[2164:] train = df[:2163]

```
data2['Date'] = df['Date']
     data2['Close'] = df['Close']
[78]: #start calculating runtime in seconds
     import time
     t = time.process_time()
[79]: #scale and reshape data
     train_set = data2.iloc[:, 1:2].values
     sc = MinMaxScaler(feature_range = (0, 1))
     training_set_scaled = sc.fit_transform(train_set)
     X train = []
     y_train = []
     for i in range(60, training_set_scaled.shape[0]):
        X_train.append(training_set_scaled[i-60:i, 0])
        y_train.append(training_set_scaled[i, 0])
     X_train, y_train = np.array(X_train), np.array(y_train)
     X_train = np.reshape(X_train, (X_train.shape[0], X_train.shape[1], 1))
[80]: #add layers
     regressor = Sequential()
     regressor.add(LSTM(units = 50, return_sequences = True, input_shape = (X_train.
     \rightarrowshape[1], 1)))
     regressor.add(Dropout(0.15))
     regressor.add(LSTM(units = 50, return_sequences = True))
     regressor.add(Dropout(0.15))
     regressor.add(LSTM(units = 50, return_sequences = True))
     regressor.add(Dropout(0.15))
     regressor.add(LSTM(units = 50))
     regressor.add(Dropout(0.15))
     regressor.add(Dense(units = 1))
[81]: #add optimizer and build model
     regressor.compile(optimizer = 'adam', loss = 'mean_squared_error')
     regressor.fit(X_train, y_train, epochs = 1000, batch_size = 80)
    Epoch 1/1000
    Epoch 2/1000
    Epoch 3/1000
    34/34 [============= ] - 5s 151ms/step - loss: 0.0017
    Epoch 4/1000
    34/34 [============= ] - 5s 151ms/step - loss: 0.0015
    Epoch 5/1000
```

```
Epoch 6/1000
Epoch 7/1000
Epoch 8/1000
34/34 [============= ] - 5s 152ms/step - loss: 0.0016
Epoch 9/1000
34/34 [================= ] - 5s 151ms/step - loss: 0.0015
Epoch 10/1000
34/34 [================= ] - 5s 152ms/step - loss: 0.0013
Epoch 11/1000
34/34 [============== ] - 5s 152ms/step - loss: 0.0013
Epoch 12/1000
Epoch 13/1000
34/34 [============== ] - 5s 152ms/step - loss: 0.0013
Epoch 14/1000
34/34 [============== ] - 5s 151ms/step - loss: 0.0013
Epoch 15/1000
34/34 [============= ] - 5s 151ms/step - loss: 0.0011
Epoch 16/1000
Epoch 17/1000
34/34 [============= ] - 5s 152ms/step - loss: 0.0011
Epoch 18/1000
34/34 [============ ] - 5s 151ms/step - loss: 0.0012
Epoch 19/1000
34/34 [============== ] - 5s 150ms/step - loss: 9.5887e-04
Epoch 20/1000
34/34 [=============== ] - 5s 152ms/step - loss: 9.6538e-04
Epoch 21/1000
34/34 [============== ] - 5s 152ms/step - loss: 0.0016
Epoch 22/1000
Epoch 23/1000
Epoch 24/1000
34/34 [================= ] - 5s 151ms/step - loss: 0.0011
Epoch 25/1000
Epoch 26/1000
34/34 [============ ] - 5s 148ms/step - loss: 0.0012
Epoch 27/1000
34/34 [=============== ] - 5s 151ms/step - loss: 9.0647e-04
Epoch 28/1000
34/34 [=============== ] - 5s 150ms/step - loss: 8.6321e-04
Epoch 29/1000
```

```
Epoch 30/1000
Epoch 31/1000
34/34 [=============== ] - 5s 151ms/step - loss: 9.7050e-04
Epoch 32/1000
34/34 [============= ] - 5s 148ms/step - loss: 0.0012
Epoch 33/1000
34/34 [================= ] - 5s 148ms/step - loss: 8.2260e-04
Epoch 34/1000
34/34 [================== ] - 5s 148ms/step - loss: 8.5902e-04
Epoch 35/1000
34/34 [================ ] - 5s 151ms/step - loss: 8.6987e-04
Epoch 36/1000
34/34 [============== ] - 5s 148ms/step - loss: 7.8306e-04
Epoch 37/1000
34/34 [=============== ] - 5s 147ms/step - loss: 7.9846e-04
Epoch 38/1000
34/34 [=============== ] - 5s 149ms/step - loss: 7.7654e-04
Epoch 39/1000
34/34 [============== ] - 5s 148ms/step - loss: 6.7745e-04
Epoch 40/1000
Epoch 41/1000
34/34 [=============== ] - 5s 148ms/step - loss: 7.9105e-04
Epoch 42/1000
34/34 [=============== ] - 5s 149ms/step - loss: 9.8635e-04
Epoch 43/1000
34/34 [============== ] - 5s 150ms/step - loss: 8.0795e-04
Epoch 44/1000
34/34 [=============== ] - 5s 148ms/step - loss: 7.5662e-04
Epoch 45/1000
Epoch 46/1000
Epoch 47/1000
34/34 [================= ] - 5s 149ms/step - loss: 6.6932e-04
Epoch 48/1000
34/34 [================== ] - 5s 149ms/step - loss: 7.1821e-04
Epoch 49/1000
34/34 [================= ] - 5s 147ms/step - loss: 7.3150e-04
Epoch 50/1000
Epoch 51/1000
Epoch 52/1000
34/34 [============== ] - 5s 148ms/step - loss: 0.0011
Epoch 53/1000
34/34 [============= ] - 5s 147ms/step - loss: 7.3493e-04
```

```
Epoch 54/1000
Epoch 55/1000
Epoch 56/1000
34/34 [=============== ] - 5s 159ms/step - loss: 6.7048e-04
Epoch 57/1000
34/34 [================= ] - 5s 147ms/step - loss: 6.9542e-04
Epoch 58/1000
34/34 [================== ] - 5s 148ms/step - loss: 6.6462e-04
Epoch 59/1000
Epoch 60/1000
Epoch 61/1000
34/34 [=============== ] - 5s 147ms/step - loss: 6.1679e-04
Epoch 62/1000
34/34 [============== ] - 5s 146ms/step - loss: 0.0010
Epoch 63/1000
34/34 [============== ] - 5s 147ms/step - loss: 7.1299e-04
Epoch 64/1000
Epoch 65/1000
34/34 [================ ] - 5s 146ms/step - loss: 6.5916e-04
Epoch 66/1000
34/34 [================= ] - 5s 147ms/step - loss: 6.5299e-04
Epoch 67/1000
34/34 [============== ] - 5s 147ms/step - loss: 6.9246e-04
Epoch 68/1000
34/34 [=============== ] - 5s 147ms/step - loss: 5.9175e-04
Epoch 69/1000
Epoch 70/1000
Epoch 71/1000
34/34 [================ ] - 5s 146ms/step - loss: 7.1528e-04
Epoch 72/1000
34/34 [================== ] - 5s 145ms/step - loss: 6.1344e-04
Epoch 73/1000
34/34 [================== ] - 5s 147ms/step - loss: 7.3738e-04
Epoch 74/1000
34/34 [================= ] - 5s 147ms/step - loss: 6.1840e-04
Epoch 75/1000
34/34 [=============== ] - 5s 146ms/step - loss: 5.1817e-04
Epoch 76/1000
34/34 [================ ] - 5s 147ms/step - loss: 5.3819e-04
Epoch 77/1000
```

```
Epoch 78/1000
Epoch 79/1000
34/34 [=============== ] - 5s 146ms/step - loss: 5.5258e-04
Epoch 80/1000
34/34 [=============== ] - 5s 146ms/step - loss: 6.7984e-04
Epoch 81/1000
34/34 [================ ] - 5s 148ms/step - loss: 5.5905e-04
Epoch 82/1000
34/34 [================== ] - 5s 148ms/step - loss: 5.9659e-04
Epoch 83/1000
34/34 [=============== ] - 5s 146ms/step - loss: 5.0197e-04
Epoch 84/1000
Epoch 85/1000
34/34 [=============== ] - 5s 145ms/step - loss: 5.7839e-04
Epoch 86/1000
34/34 [=============== ] - 5s 144ms/step - loss: 5.0456e-04
Epoch 87/1000
34/34 [============== ] - 5s 146ms/step - loss: 5.4454e-04
Epoch 88/1000
Epoch 89/1000
34/34 [=============== ] - 5s 145ms/step - loss: 5.6392e-04
Epoch 90/1000
34/34 [=============== ] - 5s 147ms/step - loss: 5.9029e-04
Epoch 91/1000
34/34 [============== ] - 5s 145ms/step - loss: 5.1701e-04
Epoch 92/1000
34/34 [=============== ] - 5s 145ms/step - loss: 5.4413e-04
Epoch 93/1000
Epoch 94/1000
Epoch 95/1000
34/34 [================= ] - 5s 146ms/step - loss: 4.9799e-04
Epoch 96/1000
34/34 [================== ] - 5s 145ms/step - loss: 5.6198e-04
Epoch 97/1000
34/34 [================== ] - 5s 145ms/step - loss: 5.2710e-04
Epoch 98/1000
Epoch 99/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.9484e-04
Epoch 100/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.7913e-04
Epoch 101/1000
```

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Epoch 102/1000
Epoch 103/1000
34/34 [================= ] - 5s 147ms/step - loss: 5.4352e-04
Epoch 104/1000
34/34 [=============== ] - 5s 153ms/step - loss: 5.3517e-04
Epoch 105/1000
34/34 [================= ] - 6s 167ms/step - loss: 4.8087e-04
Epoch 106/1000
34/34 [================== ] - 5s 145ms/step - loss: 5.2159e-04
Epoch 107/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.5120e-04
Epoch 108/1000
Epoch 109/1000
34/34 [=============== ] - 5s 146ms/step - loss: 4.9398e-04
Epoch 110/1000
34/34 [================ ] - 5s 143ms/step - loss: 5.1579e-04
Epoch 111/1000
Epoch 112/1000
Epoch 113/1000
34/34 [================ ] - 5s 146ms/step - loss: 5.6988e-04
Epoch 114/1000
Epoch 115/1000
Epoch 116/1000
34/34 [=============== ] - 6s 165ms/step - loss: 5.1234e-04
Epoch 117/1000
Epoch 118/1000
34/34 [============= ] - 5s 146ms/step - loss: 4.6679e-04
Epoch 119/1000
34/34 [================ ] - 5s 145ms/step - loss: 5.3693e-04
Epoch 120/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.8845e-04
Epoch 121/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.5954e-04
Epoch 122/1000
34/34 [=============== ] - 5s 147ms/step - loss: 5.0753e-04
Epoch 123/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.9010e-04
Epoch 124/1000
34/34 [=============== ] - 5s 146ms/step - loss: 5.3392e-04
Epoch 125/1000
34/34 [============== ] - 5s 147ms/step - loss: 4.7685e-04
```

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Epoch 126/1000
Epoch 127/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.6501e-04
Epoch 128/1000
34/34 [=============== ] - 5s 147ms/step - loss: 4.3924e-04
Epoch 129/1000
34/34 [================== ] - 5s 144ms/step - loss: 4.8060e-04
Epoch 130/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.1034e-04
Epoch 131/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.6734e-04
Epoch 132/1000
34/34 [============= ] - 5s 147ms/step - loss: 4.4904e-04
Epoch 133/1000
34/34 [=============== ] - 5s 147ms/step - loss: 3.9306e-04
Epoch 134/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9569e-04
Epoch 135/1000
Epoch 136/1000
Epoch 137/1000
34/34 [============== ] - 5s 142ms/step - loss: 4.6617e-04
Epoch 138/1000
34/34 [=============== ] - 5s 145ms/step - loss: 5.3565e-04
Epoch 139/1000
34/34 [============= ] - 5s 143ms/step - loss: 4.4372e-04
Epoch 140/1000
34/34 [=============== ] - 5s 146ms/step - loss: 7.2141e-04
Epoch 141/1000
Epoch 142/1000
Epoch 143/1000
34/34 [================== ] - 5s 144ms/step - loss: 5.0497e-04
Epoch 144/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.3946e-04
Epoch 145/1000
34/34 [================= ] - 5s 144ms/step - loss: 4.2427e-04
Epoch 146/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.6243e-04
Epoch 147/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.7101e-04
Epoch 148/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.0195e-04
Epoch 149/1000
34/34 [============= ] - 5s 145ms/step - loss: 4.1970e-04
```

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Epoch 150/1000
Epoch 151/1000
34/34 [================ ] - 5s 145ms/step - loss: 6.3813e-04
Epoch 152/1000
34/34 [============== ] - 5s 144ms/step - loss: 5.5996e-04
Epoch 153/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.1537e-04
Epoch 154/1000
34/34 [================= ] - 5s 145ms/step - loss: 3.9184e-04
Epoch 155/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.1074e-04
Epoch 156/1000
34/34 [============= ] - 5s 143ms/step - loss: 4.0084e-04
Epoch 157/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.7691e-04
Epoch 158/1000
34/34 [=============== ] - 5s 144ms/step - loss: 5.3454e-04
Epoch 159/1000
Epoch 160/1000
Epoch 161/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.0917e-04
Epoch 162/1000
Epoch 163/1000
34/34 [============== ] - 5s 143ms/step - loss: 4.4131e-04
Epoch 164/1000
34/34 [=============== ] - 5s 157ms/step - loss: 4.0889e-04
Epoch 165/1000
34/34 [=============== ] - 5s 155ms/step - loss: 4.0094e-04
Epoch 166/1000
Epoch 167/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.8624e-04
Epoch 168/1000
34/34 [================== ] - 5s 144ms/step - loss: 4.2292e-04
Epoch 169/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.5461e-04
Epoch 170/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.7421e-04
Epoch 171/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.2396e-04
Epoch 172/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.9311e-04
Epoch 173/1000
```

```
Epoch 174/1000
Epoch 175/1000
34/34 [=============== ] - 5s 148ms/step - loss: 4.2229e-04
Epoch 176/1000
34/34 [=============== ] - 6s 171ms/step - loss: 3.9051e-04
Epoch 177/1000
34/34 [================= ] - 5s 159ms/step - loss: 3.8627e-04
Epoch 178/1000
34/34 [================== ] - 5s 143ms/step - loss: 4.7499e-04
Epoch 179/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.1292e-04
Epoch 180/1000
34/34 [============== ] - 5s 143ms/step - loss: 5.1488e-04
Epoch 181/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.5866e-04
Epoch 182/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5081e-04
Epoch 183/1000
Epoch 184/1000
Epoch 185/1000
34/34 [=============== ] - 5s 147ms/step - loss: 3.9142e-04
Epoch 186/1000
34/34 [================ ] - 5s 146ms/step - loss: 4.2567e-04
Epoch 187/1000
34/34 [============== ] - 5s 146ms/step - loss: 3.7757e-04
Epoch 188/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.8109e-04
Epoch 189/1000
34/34 [============= ] - 5s 142ms/step - loss: 4.6780e-04
Epoch 190/1000
Epoch 191/1000
34/34 [=============== ] - 5s 142ms/step - loss: 5.6482e-04
Epoch 192/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.7343e-04
Epoch 193/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.8457e-04
Epoch 194/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.4884e-04
Epoch 195/1000
Epoch 196/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.8916e-04
Epoch 197/1000
34/34 [============= ] - 5s 144ms/step - loss: 3.7386e-04
```

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Epoch 198/1000
Epoch 199/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.4538e-04
Epoch 200/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.9177e-04
Epoch 201/1000
34/34 [================ ] - 5s 142ms/step - loss: 3.6880e-04
Epoch 202/1000
34/34 [================== ] - 5s 142ms/step - loss: 3.5894e-04
Epoch 203/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.7925e-04
Epoch 204/1000
Epoch 205/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.0682e-04
Epoch 206/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3862e-04
Epoch 207/1000
Epoch 208/1000
Epoch 209/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.6231e-04
Epoch 210/1000
34/34 [=============== ] - 5s 146ms/step - loss: 4.1201e-04
Epoch 211/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.8410e-04
Epoch 212/1000
34/34 [=============== ] - 5s 143ms/step - loss: 4.2275e-04
Epoch 213/1000
Epoch 214/1000
34/34 [============== ] - 5s 144ms/step - loss: 4.0524e-04
Epoch 215/1000
34/34 [=============== ] - 5s 143ms/step - loss: 4.0319e-04
Epoch 216/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.4667e-04
Epoch 217/1000
34/34 [================== ] - 5s 145ms/step - loss: 4.5162e-04
Epoch 218/1000
34/34 [=============== ] - 5s 146ms/step - loss: 4.8240e-04
Epoch 219/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9033e-04
Epoch 220/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9805e-04
Epoch 221/1000
34/34 [=============== ] - 5s 146ms/step - loss: 5.0043e-04
```

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Epoch 222/1000
Epoch 223/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.8570e-04
Epoch 224/1000
34/34 [=============== ] - 5s 148ms/step - loss: 3.8131e-04
Epoch 225/1000
34/34 [================= ] - 6s 166ms/step - loss: 3.8944e-04
Epoch 226/1000
34/34 [================== ] - 5s 146ms/step - loss: 4.0290e-04
Epoch 227/1000
34/34 [=============== ] - 5s 146ms/step - loss: 4.2303e-04
Epoch 228/1000
Epoch 229/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.2199e-04
Epoch 230/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.3759e-04
Epoch 231/1000
Epoch 232/1000
Epoch 233/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.0728e-04
Epoch 234/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5371e-04
Epoch 235/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.7726e-04
Epoch 236/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.9825e-04
Epoch 237/1000
Epoch 238/1000
Epoch 239/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.7965e-04
Epoch 240/1000
34/34 [================== ] - 5s 144ms/step - loss: 4.0014e-04
Epoch 241/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.3152e-04
Epoch 242/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.6340e-04
Epoch 243/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.5620e-04
Epoch 244/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.6415e-04
Epoch 245/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9790e-04
```

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Epoch 246/1000
Epoch 247/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.5916e-04
Epoch 248/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.8996e-04
Epoch 249/1000
34/34 [================ ] - 5s 142ms/step - loss: 3.9650e-04
Epoch 250/1000
34/34 [================== ] - 5s 142ms/step - loss: 3.4852e-04
Epoch 251/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.3056e-04
Epoch 252/1000
Epoch 253/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.9664e-04
Epoch 254/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.6687e-04
Epoch 255/1000
Epoch 256/1000
Epoch 257/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.7054e-04
Epoch 258/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.8030e-04
Epoch 259/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.7903e-04
Epoch 260/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.6387e-04
Epoch 261/1000
Epoch 262/1000
Epoch 263/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.4945e-04
Epoch 264/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.9925e-04
Epoch 265/1000
34/34 [================== ] - 5s 142ms/step - loss: 5.8018e-04
Epoch 266/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9740e-04
Epoch 267/1000
Epoch 268/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.4352e-04
Epoch 269/1000
34/34 [=============== ] - 5s 145ms/step - loss: 4.0693e-04
```

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Epoch 270/1000
Epoch 271/1000
34/34 [================ ] - 5s 146ms/step - loss: 3.4742e-04
Epoch 272/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.7367e-04
Epoch 273/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.5659e-04
Epoch 274/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.6747e-04
Epoch 275/1000
34/34 [================ ] - 5s 146ms/step - loss: 3.4240e-04
Epoch 276/1000
34/34 [============= ] - 5s 144ms/step - loss: 4.1749e-04
Epoch 277/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.6198e-04
Epoch 278/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5654e-04
Epoch 279/1000
Epoch 280/1000
Epoch 281/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.3037e-04
Epoch 282/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3423e-04
Epoch 283/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.8679e-04
Epoch 284/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.0828e-04
Epoch 285/1000
Epoch 286/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0640e-04
Epoch 287/1000
34/34 [=============== ] - 5s 142ms/step - loss: 4.9353e-04
Epoch 288/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.5658e-04
Epoch 289/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.8122e-04
Epoch 290/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.7070e-04
Epoch 291/1000
Epoch 292/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5393e-04
Epoch 293/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5141e-04
```

```
Epoch 294/1000
Epoch 295/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1265e-04
Epoch 296/1000
34/34 [============== ] - 5s 144ms/step - loss: 4.7998e-04
Epoch 297/1000
34/34 [================= ] - 5s 141ms/step - loss: 3.6624e-04
Epoch 298/1000
34/34 [================== ] - 6s 162ms/step - loss: 3.4987e-04
Epoch 299/1000
Epoch 300/1000
Epoch 301/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.8630e-04
Epoch 302/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.6539e-04
Epoch 303/1000
Epoch 304/1000
Epoch 305/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.5695e-04
Epoch 306/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2428e-04
Epoch 307/1000
34/34 [============= ] - 5s 143ms/step - loss: 3.3704e-04
Epoch 308/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.4773e-04
Epoch 309/1000
34/34 [================ ] - 5s 141ms/step - loss: 3.9174e-04
Epoch 310/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0366e-04
Epoch 311/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1228e-04
Epoch 312/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.3098e-04
Epoch 313/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.6009e-04
Epoch 314/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.6244e-04
Epoch 315/1000
Epoch 316/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.8126e-04
Epoch 317/1000
```

```
Epoch 318/1000
Epoch 319/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4076e-04
Epoch 320/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.5298e-04
Epoch 321/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.5294e-04
Epoch 322/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.4011e-04
Epoch 323/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3063e-04
Epoch 324/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.3812e-04
Epoch 325/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.4963e-04
Epoch 326/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.4569e-04
Epoch 327/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.5424e-04
Epoch 328/1000
Epoch 329/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.6643e-04
Epoch 330/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.7657e-04
Epoch 331/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.2168e-04
Epoch 332/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5892e-04
Epoch 333/1000
Epoch 334/1000
Epoch 335/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2645e-04
Epoch 336/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.3294e-04
Epoch 337/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.3698e-04
Epoch 338/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.2424e-04
Epoch 339/1000
Epoch 340/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0806e-04
Epoch 341/1000
```

```
Epoch 342/1000
Epoch 343/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4804e-04
Epoch 344/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3648e-04
Epoch 345/1000
34/34 [================= ] - 5s 156ms/step - loss: 3.6497e-04
Epoch 346/1000
34/34 [================= ] - 5s 154ms/step - loss: 3.3633e-04
Epoch 347/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1427e-04
Epoch 348/1000
34/34 [============== ] - 5s 141ms/step - loss: 3.4620e-04
Epoch 349/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3523e-04
Epoch 350/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.7005e-04
Epoch 351/1000
Epoch 352/1000
Epoch 353/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.7357e-04
Epoch 354/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3051e-04
Epoch 355/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.1027e-04
Epoch 356/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.8002e-04
Epoch 357/1000
Epoch 358/1000
Epoch 359/1000
34/34 [=============== ] - 5s 151ms/step - loss: 4.6301e-04
Epoch 360/1000
34/34 [================== ] - 6s 169ms/step - loss: 3.1300e-04
Epoch 361/1000
Epoch 362/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.7727e-04
Epoch 363/1000
Epoch 364/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.4611e-04
Epoch 365/1000
34/34 [============] - 5s 143ms/step - loss: 3.4017e-04
```

```
Epoch 366/1000
Epoch 367/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1734e-04
Epoch 368/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.1946e-04
Epoch 369/1000
34/34 [================ ] - 5s 143ms/step - loss: 3.7246e-04
Epoch 370/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.6759e-04
Epoch 371/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.7261e-04
Epoch 372/1000
Epoch 373/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0721e-04
Epoch 374/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.2171e-04
Epoch 375/1000
Epoch 376/1000
Epoch 377/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.6881e-04
Epoch 378/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3085e-04
Epoch 379/1000
Epoch 380/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3919e-04
Epoch 381/1000
Epoch 382/1000
34/34 [============== ] - 5s 146ms/step - loss: 3.2319e-04
Epoch 383/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.5063e-04
Epoch 384/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.6918e-04
Epoch 385/1000
34/34 [================== ] - 5s 144ms/step - loss: 4.0705e-04
Epoch 386/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.7935e-04
Epoch 387/1000
Epoch 388/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.1390e-04
Epoch 389/1000
34/34 [============= ] - 5s 144ms/step - loss: 3.0360e-04
```

```
Epoch 390/1000
Epoch 391/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0408e-04
Epoch 392/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0041e-04
Epoch 393/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.3184e-04
Epoch 394/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.4863e-04
Epoch 395/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.1613e-04
Epoch 396/1000
Epoch 397/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.6341e-04
Epoch 398/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0316e-04
Epoch 399/1000
Epoch 400/1000
Epoch 401/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4512e-04
Epoch 402/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1496e-04
Epoch 403/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.2183e-04
Epoch 404/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.8055e-04
Epoch 405/1000
Epoch 406/1000
Epoch 407/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.4057e-04
Epoch 408/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.4064e-04
Epoch 409/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.0823e-04
Epoch 410/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.9213e-04
Epoch 411/1000
34/34 [============== ] - 5s 145ms/step - loss: 3.3744e-04
Epoch 412/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.4444e-04
Epoch 413/1000
34/34 [============= ] - 5s 144ms/step - loss: 3.0440e-04
```

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Epoch 414/1000
Epoch 415/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.4794e-04
Epoch 416/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0526e-04
Epoch 417/1000
34/34 [================ ] - 5s 142ms/step - loss: 3.3277e-04
Epoch 418/1000
34/34 [================= ] - 5s 144ms/step - loss: 5.0581e-04
Epoch 419/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.2606e-04
Epoch 420/1000
Epoch 421/1000
34/34 [=============== ] - 6s 168ms/step - loss: 3.1104e-04
Epoch 422/1000
34/34 [=============== ] - 6s 164ms/step - loss: 3.0820e-04
Epoch 423/1000
Epoch 424/1000
Epoch 425/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.2085e-04
Epoch 426/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.2593e-04
Epoch 427/1000
34/34 [=============== ] - 5s 143ms/step - loss: 5.6393e-04
Epoch 428/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3726e-04
Epoch 429/1000
Epoch 430/1000
Epoch 431/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3865e-04
Epoch 432/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.3605e-04
Epoch 433/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.5590e-04
Epoch 434/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.4358e-04
Epoch 435/1000
Epoch 436/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.9530e-04
Epoch 437/1000
```

```
Epoch 438/1000
Epoch 439/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2400e-04
Epoch 440/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.7215e-04
Epoch 441/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.5150e-04
Epoch 442/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.2273e-04
Epoch 443/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5481e-04
Epoch 444/1000
Epoch 445/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0218e-04
Epoch 446/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0383e-04
Epoch 447/1000
Epoch 448/1000
Epoch 449/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0909e-04
Epoch 450/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.9237e-04
Epoch 451/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3447e-04
Epoch 452/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.2925e-04
Epoch 453/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.4662e-04
Epoch 454/1000
Epoch 455/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0614e-04
Epoch 456/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.2435e-04
Epoch 457/1000
Epoch 458/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.6631e-04
Epoch 459/1000
Epoch 460/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1785e-04
Epoch 461/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0384e-04
```

```
Epoch 462/1000
Epoch 463/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.4097e-04
Epoch 464/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4639e-04
Epoch 465/1000
34/34 [================ ] - 5s 142ms/step - loss: 3.5329e-04
Epoch 466/1000
34/34 [================== ] - 6s 163ms/step - loss: 3.5180e-04
Epoch 467/1000
34/34 [=============== ] - 5s 149ms/step - loss: 3.3324e-04
Epoch 468/1000
34/34 [============== ] - 5s 144ms/step - loss: 4.6892e-04
Epoch 469/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2528e-04
Epoch 470/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0434e-04
Epoch 471/1000
Epoch 472/1000
Epoch 473/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.5597e-04
Epoch 474/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4517e-04
Epoch 475/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4297e-04
Epoch 476/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.5922e-04
Epoch 477/1000
Epoch 478/1000
Epoch 479/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9689e-04
Epoch 480/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.0402e-04
Epoch 481/1000
34/34 [=================== ] - 5s 145ms/step - loss: 3.2344e-04
Epoch 482/1000
34/34 [=============== ] - 6s 164ms/step - loss: 3.3342e-04
Epoch 483/1000
34/34 [=============== ] - 6s 171ms/step - loss: 3.9264e-04
Epoch 484/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.2049e-04
Epoch 485/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.8358e-04
```

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Epoch 486/1000
Epoch 487/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1029e-04
Epoch 488/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4254e-04
Epoch 489/1000
34/34 [================ ] - 5s 142ms/step - loss: 3.0711e-04
Epoch 490/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.7789e-04
Epoch 491/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.2031e-04
Epoch 492/1000
34/34 [============== ] - 5s 145ms/step - loss: 2.9988e-04
Epoch 493/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2748e-04
Epoch 494/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.1763e-04
Epoch 495/1000
Epoch 496/1000
Epoch 497/1000
34/34 [=============== ] - 5s 143ms/step - loss: 4.6246e-04
Epoch 498/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.6354e-04
Epoch 499/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5177e-04
Epoch 500/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0617e-04
Epoch 501/1000
34/34 [================== ] - 5s 141ms/step - loss: 3.4205e-04
Epoch 502/1000
Epoch 503/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1496e-04
Epoch 504/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.0767e-04
Epoch 505/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.0712e-04
Epoch 506/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1655e-04
Epoch 507/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0517e-04
Epoch 508/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2285e-04
Epoch 509/1000
34/34 [============= ] - 5s 144ms/step - loss: 3.3624e-04
```

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Epoch 510/1000
Epoch 511/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.6508e-04
Epoch 512/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.2680e-04
Epoch 513/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.1428e-04
Epoch 514/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.1594e-04
Epoch 515/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.9303e-04
Epoch 516/1000
Epoch 517/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.5206e-04
Epoch 518/1000
34/34 [================ ] - 5s 145ms/step - loss: 3.1904e-04
Epoch 519/1000
Epoch 520/1000
Epoch 521/1000
34/34 [=============== ] - 5s 144ms/step - loss: 4.3571e-04
Epoch 522/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3549e-04
Epoch 523/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.4779e-04
Epoch 524/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.8687e-04
Epoch 525/1000
34/34 [================= ] - 5s 141ms/step - loss: 3.1719e-04
Epoch 526/1000
Epoch 527/1000
34/34 [=============== ] - 5s 159ms/step - loss: 3.3622e-04
Epoch 528/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.3297e-04
Epoch 529/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.9365e-04
Epoch 530/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.7023e-04
Epoch 531/1000
Epoch 532/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1951e-04
Epoch 533/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.4656e-04
```

```
Epoch 534/1000
Epoch 535/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1600e-04
Epoch 536/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1741e-04
Epoch 537/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.2914e-04
Epoch 538/1000
34/34 [================== ] - 5s 141ms/step - loss: 3.1384e-04
Epoch 539/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.2752e-04
Epoch 540/1000
Epoch 541/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8553e-04
Epoch 542/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1983e-04
Epoch 543/1000
Epoch 544/1000
34/34 [================= ] - 6s 167ms/step - loss: 3.2422e-04
Epoch 545/1000
34/34 [================= ] - 5s 158ms/step - loss: 2.8044e-04
Epoch 546/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8599e-04
Epoch 547/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.0813e-04
Epoch 548/1000
34/34 [=============== ] - 5s 147ms/step - loss: 2.7641e-04
Epoch 549/1000
34/34 [================ ] - 5s 152ms/step - loss: 4.2662e-04
Epoch 550/1000
Epoch 551/1000
34/34 [=============== ] - 5s 146ms/step - loss: 4.5123e-04
Epoch 552/1000
34/34 [================== ] - 5s 152ms/step - loss: 3.1236e-04
Epoch 553/1000
34/34 [================= ] - 5s 152ms/step - loss: 3.3302e-04
Epoch 554/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2783e-04
Epoch 555/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0695e-04
Epoch 556/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2531e-04
Epoch 557/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.9395e-04
```

```
Epoch 558/1000
Epoch 559/1000
34/34 [=============== ] - 5s 140ms/step - loss: 3.0368e-04
Epoch 560/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.3749e-04
Epoch 561/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.0455e-04
Epoch 562/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.9679e-04
Epoch 563/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9391e-04
Epoch 564/1000
Epoch 565/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1563e-04
Epoch 566/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.9464e-04
Epoch 567/1000
Epoch 568/1000
Epoch 569/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0970e-04
Epoch 570/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3986e-04
Epoch 571/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.2720e-04
Epoch 572/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0801e-04
Epoch 573/1000
Epoch 574/1000
Epoch 575/1000
34/34 [================= ] - 5s 144ms/step - loss: 2.7277e-04
Epoch 576/1000
34/34 [================== ] - 5s 145ms/step - loss: 2.8383e-04
Epoch 577/1000
34/34 [================= ] - 5s 151ms/step - loss: 3.0171e-04
Epoch 578/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5786e-04
Epoch 579/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3271e-04
Epoch 580/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3892e-04
Epoch 581/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2687e-04
```

```
Epoch 582/1000
Epoch 583/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.1596e-04
Epoch 584/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0715e-04
Epoch 585/1000
34/34 [================ ] - 5s 146ms/step - loss: 3.0913e-04
Epoch 586/1000
34/34 [================== ] - 5s 148ms/step - loss: 3.1663e-04
Epoch 587/1000
Epoch 588/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0297e-04
Epoch 589/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3218e-04
Epoch 590/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.0557e-04
Epoch 591/1000
Epoch 592/1000
Epoch 593/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9344e-04
Epoch 594/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3093e-04
Epoch 595/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3019e-04
Epoch 596/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9318e-04
Epoch 597/1000
Epoch 598/1000
Epoch 599/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.3266e-04
Epoch 600/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.2767e-04
Epoch 601/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.1145e-04
Epoch 602/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0943e-04
Epoch 603/1000
Epoch 604/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3746e-04
Epoch 605/1000
```

```
Epoch 606/1000
Epoch 607/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5078e-04
Epoch 608/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.2286e-04
Epoch 609/1000
34/34 [================= ] - 5s 145ms/step - loss: 3.2275e-04
Epoch 610/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.0965e-04
Epoch 611/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3459e-04
Epoch 612/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.1674e-04
Epoch 613/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0060e-04
Epoch 614/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.1011e-04
Epoch 615/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0800e-04
Epoch 616/1000
Epoch 617/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2766e-04
Epoch 618/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1731e-04
Epoch 619/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.4861e-04
Epoch 620/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1144e-04
Epoch 621/1000
Epoch 622/1000
Epoch 623/1000
34/34 [================= ] - 5s 144ms/step - loss: 2.8959e-04
Epoch 624/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.0911e-04
Epoch 625/1000
34/34 [================== ] - 5s 145ms/step - loss: 3.0900e-04
Epoch 626/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.0156e-04
Epoch 627/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0193e-04
Epoch 628/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5268e-04
Epoch 629/1000
```

```
Epoch 630/1000
Epoch 631/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0126e-04
Epoch 632/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.3473e-04
Epoch 633/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.0950e-04
Epoch 634/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.8990e-04
Epoch 635/1000
Epoch 636/1000
Epoch 637/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9672e-04
Epoch 638/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3338e-04
Epoch 639/1000
34/34 [============== ] - 5s 143ms/step - loss: 2.9116e-04
Epoch 640/1000
Epoch 641/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1009e-04
Epoch 642/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0105e-04
Epoch 643/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.0868e-04
Epoch 644/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1948e-04
Epoch 645/1000
Epoch 646/1000
Epoch 647/1000
34/34 [================ ] - 5s 161ms/step - loss: 3.1133e-04
Epoch 648/1000
34/34 [================== ] - 5s 153ms/step - loss: 2.7425e-04
Epoch 649/1000
34/34 [=================== ] - 5s 145ms/step - loss: 2.9745e-04
Epoch 650/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2985e-04
Epoch 651/1000
34/34 [============== ] - 5s 143ms/step - loss: 2.9101e-04
Epoch 652/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.6411e-04
Epoch 653/1000
```

```
Epoch 654/1000
Epoch 655/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4612e-04
Epoch 656/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1365e-04
Epoch 657/1000
34/34 [================= ] - 5s 145ms/step - loss: 3.2739e-04
Epoch 658/1000
34/34 [================== ] - 5s 145ms/step - loss: 5.2745e-04
Epoch 659/1000
34/34 [=============== ] - 5s 143ms/step - loss: 4.2215e-04
Epoch 660/1000
34/34 [============== ] - 5s 146ms/step - loss: 2.7916e-04
Epoch 661/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9574e-04
Epoch 662/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8751e-04
Epoch 663/1000
Epoch 664/1000
Epoch 665/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8553e-04
Epoch 666/1000
Epoch 667/1000
34/34 [============== ] - 6s 169ms/step - loss: 3.2067e-04
Epoch 668/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3945e-04
Epoch 669/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0058e-04
Epoch 670/1000
Epoch 671/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9935e-04
Epoch 672/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.2082e-04
Epoch 673/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.1960e-04
Epoch 674/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.5924e-04
Epoch 675/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.6407e-04
Epoch 676/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3315e-04
Epoch 677/1000
```

```
Epoch 678/1000
Epoch 679/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0892e-04
Epoch 680/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.8464e-04
Epoch 681/1000
34/34 [================= ] - 5s 143ms/step - loss: 2.8818e-04
Epoch 682/1000
34/34 [================== ] - 5s 142ms/step - loss: 3.1743e-04
Epoch 683/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9070e-04
Epoch 684/1000
34/34 [============== ] - 5s 144ms/step - loss: 2.9332e-04
Epoch 685/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.1692e-04
Epoch 686/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8014e-04
Epoch 687/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0327e-04
Epoch 688/1000
Epoch 689/1000
34/34 [=============== ] - 5s 143ms/step - loss: 4.0558e-04
Epoch 690/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9669e-04
Epoch 691/1000
34/34 [============= ] - 5s 143ms/step - loss: 3.4609e-04
Epoch 692/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0170e-04
Epoch 693/1000
Epoch 694/1000
Epoch 695/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.7542e-04
Epoch 696/1000
34/34 [=================== ] - 5s 145ms/step - loss: 2.8936e-04
Epoch 697/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.3805e-04
Epoch 698/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8975e-04
Epoch 699/1000
Epoch 700/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8271e-04
Epoch 701/1000
34/34 [============= ] - 5s 142ms/step - loss: 3.0797e-04
```

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Epoch 702/1000
Epoch 703/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0567e-04
Epoch 704/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0104e-04
Epoch 705/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.1438e-04
Epoch 706/1000
34/34 [================== ] - 5s 144ms/step - loss: 3.1928e-04
Epoch 707/1000
34/34 [=============== ] - 5s 151ms/step - loss: 3.6783e-04
Epoch 708/1000
Epoch 709/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1486e-04
Epoch 710/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7089e-04
Epoch 711/1000
34/34 [============== ] - 5s 141ms/step - loss: 3.4543e-04
Epoch 712/1000
Epoch 713/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0072e-04
Epoch 714/1000
Epoch 715/1000
34/34 [============== ] - 5s 142ms/step - loss: 2.9054e-04
Epoch 716/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.9004e-04
Epoch 717/1000
Epoch 718/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.5671e-04
Epoch 719/1000
34/34 [================= ] - 5s 145ms/step - loss: 3.6370e-04
Epoch 720/1000
34/34 [================== ] - 5s 144ms/step - loss: 2.8962e-04
Epoch 721/1000
Epoch 722/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.8514e-04
Epoch 723/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.0565e-04
Epoch 724/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0468e-04
Epoch 725/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9356e-04
```

```
Epoch 726/1000
Epoch 727/1000
34/34 [=============== ] - 5s 150ms/step - loss: 3.4416e-04
Epoch 728/1000
34/34 [=============== ] - 6s 169ms/step - loss: 3.2851e-04
Epoch 729/1000
34/34 [================= ] - 5s 155ms/step - loss: 2.9955e-04
Epoch 730/1000
34/34 [================== ] - 5s 144ms/step - loss: 2.7046e-04
Epoch 731/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.3348e-04
Epoch 732/1000
34/34 [============== ] - 5s 141ms/step - loss: 3.4740e-04
Epoch 733/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9710e-04
Epoch 734/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1604e-04
Epoch 735/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0723e-04
Epoch 736/1000
Epoch 737/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1133e-04
Epoch 738/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7736e-04
Epoch 739/1000
34/34 [============== ] - 5s 143ms/step - loss: 2.7859e-04
Epoch 740/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9762e-04
Epoch 741/1000
Epoch 742/1000
Epoch 743/1000
34/34 [================= ] - 5s 143ms/step - loss: 2.8345e-04
Epoch 744/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.0758e-04
Epoch 745/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.8339e-04
Epoch 746/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.9397e-04
Epoch 747/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.7454e-04
Epoch 748/1000
34/34 [============== ] - 5s 141ms/step - loss: 3.2362e-04
Epoch 749/1000
```

```
Epoch 750/1000
Epoch 751/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.0476e-04
Epoch 752/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.5073e-04
Epoch 753/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.2504e-04
Epoch 754/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.8823e-04
Epoch 755/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8441e-04
Epoch 756/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.1062e-04
Epoch 757/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.4561e-04
Epoch 758/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2312e-04
Epoch 759/1000
Epoch 760/1000
Epoch 761/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8130e-04
Epoch 762/1000
34/34 [================ ] - 5s 146ms/step - loss: 2.9847e-04
Epoch 763/1000
34/34 [============== ] - 5s 146ms/step - loss: 2.7730e-04
Epoch 764/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8358e-04
Epoch 765/1000
Epoch 766/1000
Epoch 767/1000
34/34 [================= ] - 5s 145ms/step - loss: 2.8809e-04
Epoch 768/1000
34/34 [================== ] - 5s 160ms/step - loss: 2.9844e-04
Epoch 769/1000
34/34 [================== ] - 5s 150ms/step - loss: 2.6328e-04
Epoch 770/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9280e-04
Epoch 771/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0154e-04
Epoch 772/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2230e-04
Epoch 773/1000
```

```
Epoch 774/1000
Epoch 775/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.2739e-04
Epoch 776/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9951e-04
Epoch 777/1000
34/34 [================= ] - 5s 145ms/step - loss: 2.9773e-04
Epoch 778/1000
34/34 [=================== ] - 5s 142ms/step - loss: 2.8302e-04
Epoch 779/1000
Epoch 780/1000
34/34 [============= ] - 5s 141ms/step - loss: 2.9028e-04
Epoch 781/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9408e-04
Epoch 782/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.3898e-04
Epoch 783/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.0376e-04
Epoch 784/1000
Epoch 785/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.2380e-04
Epoch 786/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.6816e-04
Epoch 787/1000
34/34 [============== ] - 5s 144ms/step - loss: 2.6224e-04
Epoch 788/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0434e-04
Epoch 789/1000
Epoch 790/1000
Epoch 791/1000
34/34 [================ ] - 5s 143ms/step - loss: 3.1237e-04
Epoch 792/1000
34/34 [================== ] - 5s 144ms/step - loss: 2.8417e-04
Epoch 793/1000
Epoch 794/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7722e-04
Epoch 795/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9300e-04
Epoch 796/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.3394e-04
Epoch 797/1000
```

```
Epoch 798/1000
Epoch 799/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7727e-04
Epoch 800/1000
34/34 [================ ] - 5s 142ms/step - loss: 2.8894e-04
Epoch 801/1000
34/34 [================= ] - 5s 142ms/step - loss: 2.7228e-04
Epoch 802/1000
34/34 [=================== ] - 5s 145ms/step - loss: 3.0995e-04
Epoch 803/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.2305e-04
Epoch 804/1000
Epoch 805/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.3824e-04
Epoch 806/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1293e-04
Epoch 807/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.3800e-04
Epoch 808/1000
34/34 [================= ] - 5s 141ms/step - loss: 2.9411e-04
Epoch 809/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.6109e-04
Epoch 810/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0893e-04
Epoch 811/1000
34/34 [============= ] - 5s 143ms/step - loss: 2.8449e-04
Epoch 812/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7317e-04
Epoch 813/1000
Epoch 814/1000
Epoch 815/1000
34/34 [================= ] - 5s 142ms/step - loss: 2.9579e-04
Epoch 816/1000
34/34 [=================== ] - 5s 145ms/step - loss: 3.0889e-04
Epoch 817/1000
34/34 [================= ] - 5s 144ms/step - loss: 3.0182e-04
Epoch 818/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.8451e-04
Epoch 819/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.0592e-04
Epoch 820/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9380e-04
Epoch 821/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0396e-04
```

```
Epoch 822/1000
Epoch 823/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.7971e-04
Epoch 824/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2278e-04
Epoch 825/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.1757e-04
Epoch 826/1000
Epoch 827/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.9584e-04
Epoch 828/1000
34/34 [============= ] - 5s 147ms/step - loss: 3.0081e-04
Epoch 829/1000
34/34 [=============== ] - 6s 165ms/step - loss: 3.1260e-04
Epoch 830/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.1373e-04
Epoch 831/1000
Epoch 832/1000
Epoch 833/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2919e-04
Epoch 834/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7810e-04
Epoch 835/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8198e-04
Epoch 836/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1437e-04
Epoch 837/1000
Epoch 838/1000
Epoch 839/1000
34/34 [=============== ] - 5s 141ms/step - loss: 3.1506e-04
Epoch 840/1000
34/34 [================== ] - 5s 142ms/step - loss: 2.8897e-04
Epoch 841/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.9931e-04
Epoch 842/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9408e-04
Epoch 843/1000
34/34 [============== ] - 5s 143ms/step - loss: 2.9530e-04
Epoch 844/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9507e-04
Epoch 845/1000
34/34 [============= ] - 5s 143ms/step - loss: 3.1339e-04
```

```
Epoch 846/1000
Epoch 847/1000
34/34 [=============== ] - 5s 142ms/step - loss: 3.3076e-04
Epoch 848/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0165e-04
Epoch 849/1000
34/34 [================= ] - 5s 143ms/step - loss: 3.0395e-04
Epoch 850/1000
34/34 [================== ] - 5s 152ms/step - loss: 2.7826e-04
Epoch 851/1000
Epoch 852/1000
Epoch 853/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9762e-04
Epoch 854/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8095e-04
Epoch 855/1000
Epoch 856/1000
Epoch 857/1000
34/34 [================= ] - 5s 143ms/step - loss: 2.8824e-04
Epoch 858/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.0264e-04
Epoch 859/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.7816e-04
Epoch 860/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.8696e-04
Epoch 861/1000
Epoch 862/1000
Epoch 863/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.7477e-04
Epoch 864/1000
34/34 [================== ] - 5s 142ms/step - loss: 3.3436e-04
Epoch 865/1000
34/34 [================== ] - 5s 141ms/step - loss: 3.1816e-04
Epoch 866/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1810e-04
Epoch 867/1000
Epoch 868/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7369e-04
Epoch 869/1000
```

```
Epoch 870/1000
Epoch 871/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.7553e-04
Epoch 872/1000
34/34 [============== ] - 5s 144ms/step - loss: 3.0575e-04
Epoch 873/1000
34/34 [================ ] - 5s 142ms/step - loss: 2.9269e-04
Epoch 874/1000
Epoch 875/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9770e-04
Epoch 876/1000
34/34 [============== ] - 5s 142ms/step - loss: 3.1020e-04
Epoch 877/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1302e-04
Epoch 878/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.1304e-04
Epoch 879/1000
Epoch 880/1000
Epoch 881/1000
34/34 [================ ] - 5s 143ms/step - loss: 2.8786e-04
Epoch 882/1000
34/34 [=============== ] - 5s 146ms/step - loss: 2.7916e-04
Epoch 883/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8496e-04
Epoch 884/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8094e-04
Epoch 885/1000
Epoch 886/1000
Epoch 887/1000
34/34 [================ ] - 5s 142ms/step - loss: 2.7976e-04
Epoch 888/1000
34/34 [================== ] - 5s 142ms/step - loss: 2.9497e-04
Epoch 889/1000
34/34 [================== ] - 5s 153ms/step - loss: 2.6755e-04
Epoch 890/1000
34/34 [=============== ] - 5s 154ms/step - loss: 2.9928e-04
Epoch 891/1000
34/34 [============== ] - 5s 146ms/step - loss: 2.9744e-04
Epoch 892/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.5676e-04
Epoch 893/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.0074e-04
```

```
Epoch 894/1000
Epoch 895/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9936e-04
Epoch 896/1000
34/34 [============= ] - 5s 142ms/step - loss: 3.2033e-04
Epoch 897/1000
34/34 [================ ] - 5s 144ms/step - loss: 2.8326e-04
Epoch 898/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.8430e-04
Epoch 899/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9435e-04
Epoch 900/1000
Epoch 901/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8010e-04
Epoch 902/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8928e-04
Epoch 903/1000
Epoch 904/1000
Epoch 905/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1139e-04
Epoch 906/1000
34/34 [================ ] - 5s 146ms/step - loss: 2.8188e-04
Epoch 907/1000
34/34 [============= ] - 5s 144ms/step - loss: 2.5378e-04
Epoch 908/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.9531e-04
Epoch 909/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9770e-04
Epoch 910/1000
34/34 [============== ] - 5s 144ms/step - loss: 2.8494e-04
Epoch 911/1000
34/34 [================ ] - 5s 146ms/step - loss: 2.9079e-04
Epoch 912/1000
Epoch 913/1000
Epoch 914/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7938e-04
Epoch 915/1000
Epoch 916/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9746e-04
Epoch 917/1000
```

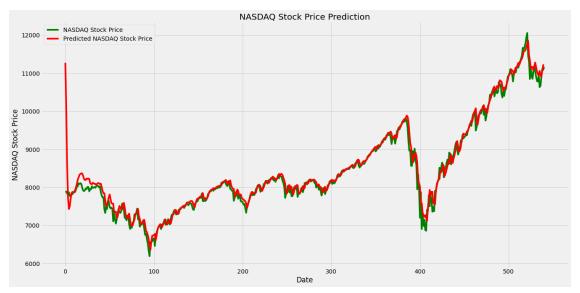
```
Epoch 918/1000
Epoch 919/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9462e-04
Epoch 920/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.1155e-04
Epoch 921/1000
34/34 [================ ] - 5s 143ms/step - loss: 3.2100e-04
Epoch 922/1000
34/34 [================= ] - 5s 142ms/step - loss: 3.3191e-04
Epoch 923/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2750e-04
Epoch 924/1000
Epoch 925/1000
34/34 [=============== ] - 5s 141ms/step - loss: 2.7730e-04
Epoch 926/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.7742e-04
Epoch 927/1000
Epoch 928/1000
Epoch 929/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.7137e-04
Epoch 930/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.7604e-04
Epoch 931/1000
34/34 [=============== ] - 5s 146ms/step - loss: 2.9068e-04
Epoch 932/1000
34/34 [=============== ] - 5s 146ms/step - loss: 2.5713e-04
Epoch 933/1000
Epoch 934/1000
Epoch 935/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.1462e-04
Epoch 936/1000
34/34 [================== ] - 5s 144ms/step - loss: 2.9207e-04
Epoch 937/1000
34/34 [================== ] - 5s 146ms/step - loss: 2.7927e-04
Epoch 938/1000
34/34 [=============== ] - 5s 147ms/step - loss: 3.0003e-04
Epoch 939/1000
34/34 [=============== ] - 5s 147ms/step - loss: 2.9245e-04
Epoch 940/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.0037e-04
Epoch 941/1000
34/34 [=============== ] - 5s 147ms/step - loss: 3.4602e-04
```

```
Epoch 942/1000
Epoch 943/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.9403e-04
Epoch 944/1000
34/34 [=============== ] - 5s 146ms/step - loss: 3.1332e-04
Epoch 945/1000
34/34 [================ ] - 5s 144ms/step - loss: 3.1982e-04
Epoch 946/1000
34/34 [================== ] - 5s 142ms/step - loss: 2.7802e-04
Epoch 947/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.7853e-04
Epoch 948/1000
34/34 [============== ] - 5s 144ms/step - loss: 2.9644e-04
Epoch 949/1000
34/34 [=============== ] - 5s 153ms/step - loss: 3.1260e-04
Epoch 950/1000
34/34 [=============== ] - 6s 164ms/step - loss: 3.0480e-04
Epoch 951/1000
Epoch 952/1000
Epoch 953/1000
34/34 [================ ] - 5s 145ms/step - loss: 2.9833e-04
Epoch 954/1000
Epoch 955/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.9128e-04
Epoch 956/1000
34/34 [=============== ] - 5s 145ms/step - loss: 3.4400e-04
Epoch 957/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.0111e-04
Epoch 958/1000
Epoch 959/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.5021e-04
Epoch 960/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.7879e-04
Epoch 961/1000
34/34 [================== ] - 5s 143ms/step - loss: 2.9116e-04
Epoch 962/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.7635e-04
Epoch 963/1000
Epoch 964/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.8211e-04
Epoch 965/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.8978e-04
```

```
Epoch 966/1000
Epoch 967/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.5964e-04
Epoch 968/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.7009e-04
Epoch 969/1000
34/34 [================= ] - 5s 143ms/step - loss: 2.9487e-04
Epoch 970/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.0303e-04
Epoch 971/1000
34/34 [=============== ] - 5s 146ms/step - loss: 2.8102e-04
Epoch 972/1000
Epoch 973/1000
34/34 [=============== ] - 6s 165ms/step - loss: 3.1510e-04
Epoch 974/1000
Epoch 975/1000
Epoch 976/1000
Epoch 977/1000
34/34 [=============== ] - 5s 142ms/step - loss: 2.6992e-04
Epoch 978/1000
34/34 [=============== ] - 5s 145ms/step - loss: 2.9733e-04
Epoch 979/1000
34/34 [============== ] - 5s 143ms/step - loss: 3.5024e-04
Epoch 980/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.1307e-04
Epoch 981/1000
34/34 [============= ] - 5s 144ms/step - loss: 3.0180e-04
Epoch 982/1000
Epoch 983/1000
34/34 [=============== ] - 5s 144ms/step - loss: 3.2839e-04
Epoch 984/1000
34/34 [================== ] - 5s 143ms/step - loss: 3.6074e-04
Epoch 985/1000
34/34 [================== ] - 5s 141ms/step - loss: 3.2822e-04
Epoch 986/1000
34/34 [=============== ] - 5s 143ms/step - loss: 3.2425e-04
Epoch 987/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.7430e-04
Epoch 988/1000
34/34 [=============== ] - 5s 144ms/step - loss: 2.6436e-04
Epoch 989/1000
34/34 [=============== ] - 5s 143ms/step - loss: 2.9409e-04
```

```
34/34 [============= ] - 5s 144ms/step - loss: 2.6230e-04
    Epoch 991/1000
    Epoch 992/1000
    34/34 [=============== ] - 5s 144ms/step - loss: 3.2134e-04
    Epoch 993/1000
    34/34 [================= ] - 5s 146ms/step - loss: 3.2331e-04
    Epoch 994/1000
    34/34 [============== ] - 5s 145ms/step - loss: 3.0097e-04
    Epoch 995/1000
    Epoch 996/1000
    34/34 [=============== ] - 5s 145ms/step - loss: 2.9308e-04
    Epoch 997/1000
    34/34 [=============== ] - 5s 142ms/step - loss: 2.9715e-04
    Epoch 998/1000
    34/34 [============= ] - 5s 143ms/step - loss: 3.3480e-04
    Epoch 999/1000
    34/34 [============== ] - 5s 144ms/step - loss: 2.6673e-04
    Epoch 1000/1000
    34/34 [============== ] - 5s 144ms/step - loss: 2.9082e-04
[81]: <tensorflow.python.keras.callbacks.History at 0x7fdf1871cad0>
[82]: #train the model
     testdataframe= test
     testdataframe['Date'] = testdataframe.index
     testdata = pd.DataFrame(columns = ['Date', 'Close'])
     testdata['Date'] = testdataframe['Date']
     testdata['Close'] = testdataframe['Close']
     real_stock_price = testdata.iloc[:, 1:2].values
     dataset total = pd.concat((data2['Close'], testdata['Close']), axis = 0)
     inputs = dataset_total[len(dataset_total) - len(testdata) - 60:].values
     inputs = inputs.reshape(-1,1)
     inputs = sc.transform(inputs)
     X \text{ test} = []
     for i in range(60, inputs.shape[0]):
        X_test.append(inputs[i-60:i, 0])
     X_test = np.array(X_test)
     X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], 1))
[83]: predicted_stock_price = regressor.predict(X_test)
     predicted_stock_price = sc.inverse_transform(predicted_stock_price)
[84]: plt.figure(figsize=(20,10))
     plt.plot(real_stock_price, color = 'green', label = 'NASDAQ Stock Price')
```

Epoch 990/1000



```
[85]: rmse_predict= np.reshape(predicted_stock_price,541)
```

```
[86]: def smape(a, f):
    return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)
    smape(test["Close"].values,rmse_predict)
```

[86]: 1.3944491323928099

```
[87]: from sklearn.metrics import mean_absolute_error
    from sklearn.metrics import mean_squared_error
    from sklearn.metrics import r2_score
    mae = mean_absolute_error(test["Close"].values, rmse_predict)
    mse = mean_squared_error(test["Close"].values, rmse_predict)
    rmse = np.sqrt(mse)

print("Results of sklearn.metrics:")
    print("MAE:", mae)
    print("MSE:", mse)
    print("RMSE:", rmse)
```

```
Results of sklearn.metrics:
    MAE: 117.05387249162432
    MSE: 55267.551367822314
    RMSE: 235.09051739239146
[88]: elapsed_time = time.process_time() - t
    print(elapsed_time)
    15092.296083494
[]: #using GRU to predict NASDAQ
[89]: import time
    t = time.process_time()
[90]: regressorGRU = Sequential()
    regressorGRU.add(GRU(units = 50, return_sequences = True, input_shape = __
     \hookrightarrow (X_train.shape[1],1)))
    regressorGRU.add(Dropout(0.15))
    regressorGRU.add(GRU(units = 50, return_sequences = True))
    regressorGRU.add(Dropout(0.15))
    regressorGRU.add(GRU(units = 50, return_sequences = True))
    regressorGRU.add(Dropout(0.15))
    regressorGRU.add(GRU(units = 50))
    regressorGRU.add(Dropout(0.15))
    regressorGRU.add(Dense(units = 1))
[91]: regressorGRU.compile(optimizer='adam',loss='mean_squared_error')
[92]: regressorGRU.fit(X_train, y_train, epochs = 1000, batch_size = 32)
    Epoch 1/1000
    83/83 [============= ] - 13s 154ms/step - loss: 0.0093
    Epoch 2/1000
    Epoch 3/1000
    Epoch 4/1000
    83/83 [============ ] - 13s 153ms/step - loss: 0.0012
    Epoch 5/1000
    Epoch 6/1000
    83/83 [============ ] - 12s 141ms/step - loss: 0.0010
    Epoch 7/1000
    83/83 [============ ] - 12s 140ms/step - loss: 9.9972e-04
    Epoch 8/1000
```

```
Epoch 9/1000
83/83 [============= ] - 12s 139ms/step - loss: 9.3036e-04
Epoch 10/1000
Epoch 11/1000
Epoch 12/1000
Epoch 13/1000
Epoch 14/1000
Epoch 15/1000
Epoch 16/1000
Epoch 17/1000
Epoch 18/1000
83/83 [============== ] - 11s 134ms/step - loss: 8.1228e-04
Epoch 19/1000
Epoch 20/1000
83/83 [============ ] - 11s 134ms/step - loss: 7.6414e-04
Epoch 21/1000
83/83 [============ ] - 11s 134ms/step - loss: 8.4814e-04
Epoch 22/1000
Epoch 23/1000
Epoch 24/1000
Epoch 25/1000
83/83 [============= ] - 11s 134ms/step - loss: 6.6359e-04
Epoch 26/1000
Epoch 27/1000
Epoch 28/1000
Epoch 29/1000
83/83 [============ ] - 11s 132ms/step - loss: 7.4384e-04
Epoch 30/1000
Epoch 31/1000
Epoch 32/1000
```

```
Epoch 33/1000
83/83 [============ ] - 11s 132ms/step - loss: 6.0583e-04
Epoch 34/1000
Epoch 35/1000
Epoch 36/1000
Epoch 37/1000
Epoch 38/1000
Epoch 39/1000
Epoch 40/1000
Epoch 41/1000
Epoch 42/1000
Epoch 43/1000
Epoch 44/1000
Epoch 45/1000
Epoch 46/1000
Epoch 47/1000
Epoch 48/1000
Epoch 49/1000
83/83 [============ ] - 11s 133ms/step - loss: 5.3293e-04
Epoch 50/1000
Epoch 51/1000
Epoch 52/1000
Epoch 53/1000
83/83 [============ ] - 11s 133ms/step - loss: 5.1140e-04
Epoch 54/1000
Epoch 55/1000
Epoch 56/1000
```

```
Epoch 57/1000
83/83 [============ ] - 11s 136ms/step - loss: 4.1903e-04
Epoch 58/1000
Epoch 59/1000
Epoch 60/1000
Epoch 61/1000
Epoch 62/1000
Epoch 63/1000
Epoch 64/1000
Epoch 65/1000
Epoch 66/1000
Epoch 67/1000
Epoch 68/1000
Epoch 69/1000
Epoch 70/1000
Epoch 71/1000
Epoch 72/1000
Epoch 73/1000
83/83 [============ ] - 11s 132ms/step - loss: 4.1183e-04
Epoch 74/1000
Epoch 75/1000
Epoch 76/1000
Epoch 77/1000
83/83 [============ ] - 11s 129ms/step - loss: 4.4243e-04
Epoch 78/1000
Epoch 79/1000
83/83 [===========] - 11s 129ms/step - loss: 4.3200e-04
Epoch 80/1000
```

```
Epoch 81/1000
83/83 [============ ] - 11s 129ms/step - loss: 4.2889e-04
Epoch 82/1000
Epoch 83/1000
Epoch 84/1000
Epoch 85/1000
Epoch 86/1000
Epoch 87/1000
Epoch 88/1000
Epoch 89/1000
Epoch 90/1000
Epoch 91/1000
Epoch 92/1000
83/83 [============= ] - 11s 128ms/step - loss: 3.8184e-04
Epoch 93/1000
Epoch 94/1000
Epoch 95/1000
Epoch 96/1000
Epoch 97/1000
83/83 [============ ] - 11s 129ms/step - loss: 4.0686e-04
Epoch 98/1000
Epoch 99/1000
Epoch 100/1000
83/83 [============ ] - 11s 128ms/step - loss: 4.0277e-04
Epoch 101/1000
Epoch 102/1000
Epoch 103/1000
83/83 [===========] - 11s 129ms/step - loss: 4.5464e-04
Epoch 104/1000
83/83 [============ ] - 11s 128ms/step - loss: 3.9042e-04
```

```
Epoch 105/1000
83/83 [============= ] - 10s 126ms/step - loss: 3.4455e-04
Epoch 106/1000
Epoch 107/1000
Epoch 108/1000
Epoch 109/1000
Epoch 110/1000
Epoch 111/1000
Epoch 112/1000
Epoch 113/1000
Epoch 114/1000
Epoch 115/1000
Epoch 116/1000
83/83 [============= ] - 11s 127ms/step - loss: 3.7441e-04
Epoch 117/1000
Epoch 118/1000
Epoch 119/1000
Epoch 120/1000
83/83 [============ ] - 10s 125ms/step - loss: 4.0911e-04
Epoch 121/1000
83/83 [============ ] - 11s 127ms/step - loss: 4.2328e-04
Epoch 122/1000
Epoch 123/1000
Epoch 124/1000
83/83 [============ ] - 10s 126ms/step - loss: 4.0256e-04
Epoch 125/1000
Epoch 126/1000
Epoch 127/1000
83/83 [============ ] - 12s 149ms/step - loss: 3.3943e-04
Epoch 128/1000
```

```
Epoch 129/1000
83/83 [============ ] - 10s 126ms/step - loss: 4.1912e-04
Epoch 130/1000
Epoch 131/1000
Epoch 132/1000
Epoch 133/1000
Epoch 134/1000
Epoch 135/1000
Epoch 136/1000
Epoch 137/1000
Epoch 138/1000
Epoch 139/1000
Epoch 140/1000
83/83 [============= ] - 10s 125ms/step - loss: 3.6485e-04
Epoch 141/1000
Epoch 142/1000
Epoch 143/1000
Epoch 144/1000
83/83 [============= ] - 10s 124ms/step - loss: 3.0916e-04
Epoch 145/1000
83/83 [============= ] - 10s 124ms/step - loss: 3.4309e-04
Epoch 146/1000
Epoch 147/1000
Epoch 148/1000
83/83 [============ ] - 10s 124ms/step - loss: 3.3592e-04
Epoch 149/1000
Epoch 150/1000
Epoch 151/1000
83/83 [============ ] - 10s 125ms/step - loss: 3.4191e-04
Epoch 152/1000
```

```
Epoch 153/1000
83/83 [============= ] - 10s 126ms/step - loss: 3.1823e-04
Epoch 154/1000
Epoch 155/1000
Epoch 156/1000
Epoch 157/1000
Epoch 158/1000
Epoch 159/1000
Epoch 160/1000
Epoch 161/1000
Epoch 162/1000
Epoch 163/1000
Epoch 164/1000
Epoch 165/1000
Epoch 166/1000
83/83 [============ ] - 10s 125ms/step - loss: 3.2216e-04
Epoch 167/1000
Epoch 168/1000
83/83 [============ ] - 10s 126ms/step - loss: 3.6144e-04
Epoch 169/1000
83/83 [============ ] - 10s 126ms/step - loss: 3.7769e-04
Epoch 170/1000
Epoch 171/1000
Epoch 172/1000
Epoch 173/1000
Epoch 174/1000
Epoch 175/1000
Epoch 176/1000
```

```
Epoch 177/1000
83/83 [============= ] - 10s 125ms/step - loss: 3.1596e-04
Epoch 178/1000
Epoch 179/1000
Epoch 180/1000
Epoch 181/1000
Epoch 182/1000
Epoch 183/1000
Epoch 184/1000
Epoch 185/1000
Epoch 186/1000
83/83 [============== ] - 11s 135ms/step - loss: 3.2231e-04
Epoch 187/1000
Epoch 188/1000
83/83 [============= ] - 10s 124ms/step - loss: 3.3724e-04
Epoch 189/1000
Epoch 190/1000
Epoch 191/1000
Epoch 192/1000
83/83 [============ ] - 10s 125ms/step - loss: 2.9677e-04
Epoch 193/1000
83/83 [============= ] - 10s 125ms/step - loss: 3.4237e-04
Epoch 194/1000
Epoch 195/1000
Epoch 196/1000
83/83 [============ ] - 10s 125ms/step - loss: 3.1561e-04
Epoch 197/1000
Epoch 198/1000
Epoch 199/1000
Epoch 200/1000
```

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Epoch 201/1000
83/83 [============= ] - 10s 125ms/step - loss: 3.0415e-04
Epoch 202/1000
Epoch 203/1000
Epoch 204/1000
Epoch 205/1000
Epoch 206/1000
Epoch 207/1000
Epoch 208/1000
Epoch 209/1000
Epoch 210/1000
83/83 [============== ] - 10s 123ms/step - loss: 3.8221e-04
Epoch 211/1000
Epoch 212/1000
83/83 [============= ] - 10s 123ms/step - loss: 3.0054e-04
Epoch 213/1000
83/83 [============ ] - 10s 124ms/step - loss: 3.1767e-04
Epoch 214/1000
Epoch 215/1000
Epoch 216/1000
83/83 [============= ] - 10s 126ms/step - loss: 3.2460e-04
Epoch 217/1000
83/83 [============= ] - 10s 124ms/step - loss: 3.8200e-04
Epoch 218/1000
Epoch 219/1000
Epoch 220/1000
83/83 [============= ] - 10s 125ms/step - loss: 3.1322e-04
Epoch 221/1000
Epoch 222/1000
Epoch 223/1000
Epoch 224/1000
```

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Epoch 225/1000
83/83 [============= ] - 10s 123ms/step - loss: 3.1009e-04
Epoch 226/1000
Epoch 227/1000
Epoch 228/1000
Epoch 229/1000
Epoch 230/1000
Epoch 231/1000
Epoch 232/1000
Epoch 233/1000
Epoch 234/1000
83/83 [============== ] - 10s 122ms/step - loss: 3.1759e-04
Epoch 235/1000
Epoch 236/1000
83/83 [============= ] - 10s 122ms/step - loss: 3.2588e-04
Epoch 237/1000
Epoch 238/1000
Epoch 239/1000
Epoch 240/1000
83/83 [============= ] - 10s 121ms/step - loss: 3.2685e-04
Epoch 241/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7979e-04
Epoch 242/1000
Epoch 243/1000
Epoch 244/1000
Epoch 245/1000
Epoch 246/1000
Epoch 247/1000
Epoch 248/1000
```

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Epoch 249/1000
83/83 [============ ] - 10s 123ms/step - loss: 3.6773e-04
Epoch 250/1000
Epoch 251/1000
Epoch 252/1000
Epoch 253/1000
Epoch 254/1000
Epoch 255/1000
Epoch 256/1000
Epoch 257/1000
Epoch 258/1000
83/83 [============== ] - 11s 135ms/step - loss: 2.7673e-04
Epoch 259/1000
Epoch 260/1000
83/83 [============= ] - 10s 124ms/step - loss: 3.1873e-04
Epoch 261/1000
Epoch 262/1000
Epoch 263/1000
Epoch 264/1000
83/83 [============= ] - 10s 122ms/step - loss: 3.2568e-04
Epoch 265/1000
83/83 [============= ] - 10s 122ms/step - loss: 3.3228e-04
Epoch 266/1000
Epoch 267/1000
Epoch 268/1000
83/83 [============= ] - 10s 122ms/step - loss: 3.0821e-04
Epoch 269/1000
Epoch 270/1000
Epoch 271/1000
Epoch 272/1000
```

```
Epoch 273/1000
83/83 [============= ] - 10s 123ms/step - loss: 2.9347e-04
Epoch 274/1000
Epoch 275/1000
Epoch 276/1000
Epoch 277/1000
Epoch 278/1000
Epoch 279/1000
83/83 [============ ] - 10s 124ms/step - loss: 3.0919e-04
Epoch 280/1000
Epoch 281/1000
Epoch 282/1000
83/83 [============== ] - 10s 122ms/step - loss: 3.1101e-04
Epoch 283/1000
Epoch 284/1000
83/83 [============ ] - 10s 123ms/step - loss: 3.1431e-04
Epoch 285/1000
Epoch 286/1000
Epoch 287/1000
Epoch 288/1000
Epoch 289/1000
83/83 [============ ] - 10s 123ms/step - loss: 3.1468e-04
Epoch 290/1000
Epoch 291/1000
Epoch 292/1000
Epoch 293/1000
Epoch 294/1000
Epoch 295/1000
Epoch 296/1000
```

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Epoch 297/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.9115e-04
Epoch 298/1000
Epoch 299/1000
Epoch 300/1000
Epoch 301/1000
Epoch 302/1000
Epoch 303/1000
Epoch 304/1000
Epoch 305/1000
Epoch 306/1000
Epoch 307/1000
Epoch 308/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.8307e-04
Epoch 309/1000
Epoch 310/1000
Epoch 311/1000
Epoch 312/1000
Epoch 313/1000
83/83 [============ ] - 10s 122ms/step - loss: 3.0728e-04
Epoch 314/1000
Epoch 315/1000
Epoch 316/1000
Epoch 317/1000
Epoch 318/1000
Epoch 319/1000
Epoch 320/1000
83/83 [============ ] - 10s 121ms/step - loss: 3.2007e-04
```

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Epoch 321/1000
83/83 [============= ] - 10s 120ms/step - loss: 3.0500e-04
Epoch 322/1000
Epoch 323/1000
Epoch 324/1000
Epoch 325/1000
Epoch 326/1000
Epoch 327/1000
Epoch 328/1000
Epoch 329/1000
Epoch 330/1000
Epoch 331/1000
Epoch 332/1000
83/83 [============= ] - 10s 123ms/step - loss: 3.2481e-04
Epoch 333/1000
83/83 [============ ] - 10s 121ms/step - loss: 3.0718e-04
Epoch 334/1000
Epoch 335/1000
Epoch 336/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.8208e-04
Epoch 337/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.8912e-04
Epoch 338/1000
Epoch 339/1000
Epoch 340/1000
Epoch 341/1000
Epoch 342/1000
Epoch 343/1000
Epoch 344/1000
```

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Epoch 345/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7031e-04
Epoch 346/1000
Epoch 347/1000
Epoch 348/1000
Epoch 349/1000
Epoch 350/1000
Epoch 351/1000
Epoch 352/1000
Epoch 353/1000
Epoch 354/1000
83/83 [============== ] - 10s 120ms/step - loss: 2.7323e-04
Epoch 355/1000
Epoch 356/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7251e-04
Epoch 357/1000
Epoch 358/1000
Epoch 359/1000
Epoch 360/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.7337e-04
Epoch 361/1000
83/83 [============= ] - 10s 120ms/step - loss: 3.1895e-04
Epoch 362/1000
Epoch 363/1000
Epoch 364/1000
Epoch 365/1000
Epoch 366/1000
Epoch 367/1000
Epoch 368/1000
```

```
Epoch 369/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.9231e-04
Epoch 370/1000
Epoch 371/1000
Epoch 372/1000
Epoch 373/1000
Epoch 374/1000
Epoch 375/1000
Epoch 376/1000
Epoch 377/1000
Epoch 378/1000
83/83 [============== ] - 10s 122ms/step - loss: 3.1158e-04
Epoch 379/1000
Epoch 380/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.8714e-04
Epoch 381/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.5901e-04
Epoch 382/1000
83/83 [============ ] - 10s 121ms/step - loss: 3.0143e-04
Epoch 383/1000
Epoch 384/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7766e-04
Epoch 385/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7917e-04
Epoch 386/1000
Epoch 387/1000
Epoch 388/1000
Epoch 389/1000
Epoch 390/1000
Epoch 391/1000
Epoch 392/1000
```

```
Epoch 393/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.6828e-04
Epoch 394/1000
Epoch 395/1000
Epoch 396/1000
Epoch 397/1000
Epoch 398/1000
Epoch 399/1000
Epoch 400/1000
Epoch 401/1000
Epoch 402/1000
83/83 [============== ] - 10s 121ms/step - loss: 2.9543e-04
Epoch 403/1000
Epoch 404/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.8899e-04
Epoch 405/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.8743e-04
Epoch 406/1000
Epoch 407/1000
Epoch 408/1000
83/83 [============ ] - 10s 120ms/step - loss: 3.1931e-04
Epoch 409/1000
83/83 [============ ] - 10s 121ms/step - loss: 3.0862e-04
Epoch 410/1000
Epoch 411/1000
Epoch 412/1000
Epoch 413/1000
Epoch 414/1000
Epoch 415/1000
Epoch 416/1000
```

```
Epoch 417/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.8828e-04
Epoch 418/1000
Epoch 419/1000
Epoch 420/1000
Epoch 421/1000
Epoch 422/1000
Epoch 423/1000
Epoch 424/1000
Epoch 425/1000
Epoch 426/1000
83/83 [============== ] - 11s 137ms/step - loss: 3.2613e-04
Epoch 427/1000
Epoch 428/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.8842e-04
Epoch 429/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.9537e-04
Epoch 430/1000
Epoch 431/1000
Epoch 432/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.8260e-04
Epoch 433/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.8810e-04
Epoch 434/1000
Epoch 435/1000
Epoch 436/1000
Epoch 437/1000
Epoch 438/1000
Epoch 439/1000
Epoch 440/1000
```

```
Epoch 441/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.5669e-04
Epoch 442/1000
Epoch 443/1000
Epoch 444/1000
Epoch 445/1000
Epoch 446/1000
Epoch 447/1000
Epoch 448/1000
Epoch 449/1000
Epoch 450/1000
83/83 [============== ] - 10s 121ms/step - loss: 3.1847e-04
Epoch 451/1000
Epoch 452/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.8648e-04
Epoch 453/1000
Epoch 454/1000
Epoch 455/1000
Epoch 456/1000
Epoch 457/1000
Epoch 458/1000
Epoch 459/1000
Epoch 460/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.7276e-04
Epoch 461/1000
Epoch 462/1000
Epoch 463/1000
Epoch 464/1000
```

```
Epoch 465/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.8624e-04
Epoch 466/1000
Epoch 467/1000
Epoch 468/1000
Epoch 469/1000
Epoch 470/1000
Epoch 471/1000
Epoch 472/1000
Epoch 473/1000
Epoch 474/1000
83/83 [============== ] - 10s 121ms/step - loss: 2.6933e-04
Epoch 475/1000
Epoch 476/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.9471e-04
Epoch 477/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.9874e-04
Epoch 478/1000
Epoch 479/1000
Epoch 480/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.8673e-04
Epoch 481/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7473e-04
Epoch 482/1000
Epoch 483/1000
Epoch 484/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.6383e-04
Epoch 485/1000
Epoch 486/1000
Epoch 487/1000
Epoch 488/1000
```

```
Epoch 489/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6766e-04
Epoch 490/1000
Epoch 491/1000
Epoch 492/1000
Epoch 493/1000
Epoch 494/1000
Epoch 495/1000
Epoch 496/1000
Epoch 497/1000
Epoch 498/1000
83/83 [============== ] - 10s 122ms/step - loss: 3.0313e-04
Epoch 499/1000
Epoch 500/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.8732e-04
Epoch 501/1000
83/83 [============ ] - 10s 121ms/step - loss: 3.1426e-04
Epoch 502/1000
Epoch 503/1000
Epoch 504/1000
Epoch 505/1000
83/83 [============= ] - 10s 122ms/step - loss: 2.8245e-04
Epoch 506/1000
Epoch 507/1000
Epoch 508/1000
83/83 [============= ] - 10s 122ms/step - loss: 2.8756e-04
Epoch 509/1000
Epoch 510/1000
Epoch 511/1000
Epoch 512/1000
```

```
Epoch 513/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.9414e-04
Epoch 514/1000
Epoch 515/1000
Epoch 516/1000
Epoch 517/1000
Epoch 518/1000
Epoch 519/1000
Epoch 520/1000
Epoch 521/1000
Epoch 522/1000
83/83 [============== ] - 10s 121ms/step - loss: 2.8209e-04
Epoch 523/1000
Epoch 524/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6622e-04
Epoch 525/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.9369e-04
Epoch 526/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7408e-04
Epoch 527/1000
Epoch 528/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.5388e-04
Epoch 529/1000
Epoch 530/1000
Epoch 531/1000
Epoch 532/1000
83/83 [============ ] - 10s 126ms/step - loss: 2.5692e-04
Epoch 533/1000
Epoch 534/1000
Epoch 535/1000
Epoch 536/1000
```

```
Epoch 537/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6656e-04
Epoch 538/1000
Epoch 539/1000
Epoch 540/1000
Epoch 541/1000
Epoch 542/1000
Epoch 543/1000
Epoch 544/1000
Epoch 545/1000
Epoch 546/1000
Epoch 547/1000
Epoch 548/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.7432e-04
Epoch 549/1000
Epoch 550/1000
Epoch 551/1000
Epoch 552/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.6102e-04
Epoch 553/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7791e-04
Epoch 554/1000
Epoch 555/1000
Epoch 556/1000
Epoch 557/1000
Epoch 558/1000
Epoch 559/1000
Epoch 560/1000
```

```
Epoch 561/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.7939e-04
Epoch 562/1000
Epoch 563/1000
Epoch 564/1000
Epoch 565/1000
Epoch 566/1000
Epoch 567/1000
Epoch 568/1000
Epoch 569/1000
Epoch 570/1000
83/83 [============== ] - 11s 133ms/step - loss: 2.3704e-04
Epoch 571/1000
Epoch 572/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.7019e-04
Epoch 573/1000
Epoch 574/1000
83/83 [============ ] - 10s 119ms/step - loss: 3.1253e-04
Epoch 575/1000
Epoch 576/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.8020e-04
Epoch 577/1000
Epoch 578/1000
Epoch 579/1000
Epoch 580/1000
Epoch 581/1000
Epoch 582/1000
Epoch 583/1000
Epoch 584/1000
```

```
Epoch 585/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.7485e-04
Epoch 586/1000
Epoch 587/1000
Epoch 588/1000
Epoch 589/1000
Epoch 590/1000
Epoch 591/1000
Epoch 592/1000
Epoch 593/1000
Epoch 594/1000
83/83 [============== ] - 10s 124ms/step - loss: 2.5041e-04
Epoch 595/1000
Epoch 596/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.6739e-04
Epoch 597/1000
Epoch 598/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.8294e-04
Epoch 599/1000
Epoch 600/1000
Epoch 601/1000
83/83 [============= ] - 13s 154ms/step - loss: 3.0219e-04
Epoch 602/1000
Epoch 603/1000
Epoch 604/1000
Epoch 605/1000
Epoch 606/1000
Epoch 607/1000
Epoch 608/1000
```

```
Epoch 609/1000
83/83 [============= ] - 12s 150ms/step - loss: 2.6828e-04
Epoch 610/1000
Epoch 611/1000
Epoch 612/1000
Epoch 613/1000
Epoch 614/1000
Epoch 615/1000
Epoch 616/1000
Epoch 617/1000
Epoch 618/1000
Epoch 619/1000
Epoch 620/1000
83/83 [============= ] - 12s 149ms/step - loss: 2.6725e-04
Epoch 621/1000
83/83 [============ ] - 13s 151ms/step - loss: 2.8962e-04
Epoch 622/1000
83/83 [============ ] - 13s 157ms/step - loss: 2.7987e-04
Epoch 623/1000
Epoch 624/1000
Epoch 625/1000
83/83 [============= ] - 12s 148ms/step - loss: 2.5401e-04
Epoch 626/1000
Epoch 627/1000
Epoch 628/1000
Epoch 629/1000
Epoch 630/1000
Epoch 631/1000
Epoch 632/1000
```

```
Epoch 633/1000
83/83 [============= ] - 10s 122ms/step - loss: 2.7031e-04
Epoch 634/1000
Epoch 635/1000
Epoch 636/1000
Epoch 637/1000
Epoch 638/1000
Epoch 639/1000
Epoch 640/1000
Epoch 641/1000
Epoch 642/1000
Epoch 643/1000
Epoch 644/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6260e-04
Epoch 645/1000
Epoch 646/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.6997e-04
Epoch 647/1000
Epoch 648/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7554e-04
Epoch 649/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6196e-04
Epoch 650/1000
Epoch 651/1000
Epoch 652/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7852e-04
Epoch 653/1000
Epoch 654/1000
Epoch 655/1000
Epoch 656/1000
```

```
Epoch 657/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.9407e-04
Epoch 658/1000
Epoch 659/1000
Epoch 660/1000
Epoch 661/1000
Epoch 662/1000
Epoch 663/1000
Epoch 664/1000
Epoch 665/1000
Epoch 666/1000
83/83 [============== ] - 10s 121ms/step - loss: 2.8195e-04
Epoch 667/1000
Epoch 668/1000
83/83 [============ ] - 11s 132ms/step - loss: 2.5599e-04
Epoch 669/1000
Epoch 670/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.7970e-04
Epoch 671/1000
Epoch 672/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.5685e-04
Epoch 673/1000
83/83 [============= ] - 10s 121ms/step - loss: 2.7088e-04
Epoch 674/1000
Epoch 675/1000
Epoch 676/1000
Epoch 677/1000
Epoch 678/1000
Epoch 679/1000
Epoch 680/1000
83/83 [============ ] - 10s 126ms/step - loss: 2.5825e-04
```

```
Epoch 681/1000
83/83 [============ ] - 13s 160ms/step - loss: 2.7240e-04
Epoch 682/1000
Epoch 683/1000
Epoch 684/1000
Epoch 685/1000
Epoch 686/1000
Epoch 687/1000
Epoch 688/1000
Epoch 689/1000
Epoch 690/1000
83/83 [============== ] - 10s 126ms/step - loss: 2.8231e-04
Epoch 691/1000
Epoch 692/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.7480e-04
Epoch 693/1000
Epoch 694/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.9629e-04
Epoch 695/1000
Epoch 696/1000
83/83 [============= ] - 10s 121ms/step - loss: 3.0085e-04
Epoch 697/1000
83/83 [============ ] - 11s 128ms/step - loss: 2.7676e-04
Epoch 698/1000
Epoch 699/1000
Epoch 700/1000
Epoch 701/1000
Epoch 702/1000
Epoch 703/1000
Epoch 704/1000
```

```
Epoch 705/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.5559e-04
Epoch 706/1000
Epoch 707/1000
Epoch 708/1000
Epoch 709/1000
Epoch 710/1000
Epoch 711/1000
Epoch 712/1000
Epoch 713/1000
Epoch 714/1000
Epoch 715/1000
Epoch 716/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.7247e-04
Epoch 717/1000
Epoch 718/1000
Epoch 719/1000
Epoch 720/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.4977e-04
Epoch 721/1000
83/83 [============ ] - 11s 136ms/step - loss: 2.4580e-04
Epoch 722/1000
Epoch 723/1000
Epoch 724/1000
Epoch 725/1000
Epoch 726/1000
Epoch 727/1000
Epoch 728/1000
```

```
Epoch 729/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.8932e-04
Epoch 730/1000
Epoch 731/1000
Epoch 732/1000
Epoch 733/1000
Epoch 734/1000
Epoch 735/1000
Epoch 736/1000
Epoch 737/1000
Epoch 738/1000
83/83 [============== ] - 10s 119ms/step - loss: 2.5410e-04
Epoch 739/1000
Epoch 740/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.5278e-04
Epoch 741/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.8068e-04
Epoch 742/1000
Epoch 743/1000
Epoch 744/1000
83/83 [============ ] - 10s 125ms/step - loss: 2.7894e-04
Epoch 745/1000
Epoch 746/1000
Epoch 747/1000
Epoch 748/1000
Epoch 749/1000
Epoch 750/1000
Epoch 751/1000
Epoch 752/1000
```

```
Epoch 753/1000
83/83 [============= ] - 13s 154ms/step - loss: 2.6807e-04
Epoch 754/1000
Epoch 755/1000
Epoch 756/1000
Epoch 757/1000
Epoch 758/1000
Epoch 759/1000
Epoch 760/1000
Epoch 761/1000
Epoch 762/1000
83/83 [============== ] - 10s 119ms/step - loss: 2.5729e-04
Epoch 763/1000
Epoch 764/1000
83/83 [============ ] - 10s 121ms/step - loss: 2.6758e-04
Epoch 765/1000
Epoch 766/1000
Epoch 767/1000
Epoch 768/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.5305e-04
Epoch 769/1000
83/83 [============ ] - 10s 120ms/step - loss: 2.7211e-04
Epoch 770/1000
Epoch 771/1000
Epoch 772/1000
Epoch 773/1000
Epoch 774/1000
Epoch 775/1000
Epoch 776/1000
```

```
Epoch 777/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.7673e-04
Epoch 778/1000
Epoch 779/1000
Epoch 780/1000
Epoch 781/1000
Epoch 782/1000
Epoch 783/1000
Epoch 784/1000
Epoch 785/1000
Epoch 786/1000
Epoch 787/1000
Epoch 788/1000
83/83 [============= ] - 10s 122ms/step - loss: 2.7238e-04
Epoch 789/1000
83/83 [============ ] - 12s 149ms/step - loss: 2.6147e-04
Epoch 790/1000
Epoch 791/1000
Epoch 792/1000
83/83 [============ ] - 12s 148ms/step - loss: 2.5921e-04
Epoch 793/1000
83/83 [============= ] - 12s 149ms/step - loss: 2.6505e-04
Epoch 794/1000
Epoch 795/1000
Epoch 796/1000
Epoch 797/1000
Epoch 798/1000
Epoch 799/1000
Epoch 800/1000
```

```
Epoch 801/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.9075e-04
Epoch 802/1000
Epoch 803/1000
Epoch 804/1000
Epoch 805/1000
Epoch 806/1000
Epoch 807/1000
Epoch 808/1000
Epoch 809/1000
Epoch 810/1000
83/83 [============== ] - 10s 119ms/step - loss: 2.7287e-04
Epoch 811/1000
Epoch 812/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.4939e-04
Epoch 813/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.5720e-04
Epoch 814/1000
Epoch 815/1000
Epoch 816/1000
83/83 [============= ] - 10s 120ms/step - loss: 2.9686e-04
Epoch 817/1000
Epoch 818/1000
Epoch 819/1000
Epoch 820/1000
Epoch 821/1000
Epoch 822/1000
Epoch 823/1000
Epoch 824/1000
```

```
Epoch 825/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.6053e-04
Epoch 826/1000
Epoch 827/1000
Epoch 828/1000
Epoch 829/1000
Epoch 830/1000
Epoch 831/1000
Epoch 832/1000
Epoch 833/1000
Epoch 834/1000
83/83 [============== ] - 10s 120ms/step - loss: 2.5086e-04
Epoch 835/1000
Epoch 836/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.6886e-04
Epoch 837/1000
Epoch 838/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.4324e-04
Epoch 839/1000
Epoch 840/1000
Epoch 841/1000
Epoch 842/1000
Epoch 843/1000
Epoch 844/1000
Epoch 845/1000
Epoch 846/1000
Epoch 847/1000
Epoch 848/1000
```

```
Epoch 849/1000
83/83 [============ ] - 10s 119ms/step - loss: 2.4685e-04
Epoch 850/1000
Epoch 851/1000
Epoch 852/1000
Epoch 853/1000
Epoch 854/1000
Epoch 855/1000
Epoch 856/1000
Epoch 857/1000
Epoch 858/1000
83/83 [============== ] - 10s 119ms/step - loss: 2.6531e-04
Epoch 859/1000
Epoch 860/1000
83/83 [============= ] - 10s 119ms/step - loss: 2.6595e-04
Epoch 861/1000
Epoch 862/1000
Epoch 863/1000
Epoch 864/1000
Epoch 865/1000
Epoch 866/1000
Epoch 867/1000
Epoch 868/1000
Epoch 869/1000
Epoch 870/1000
Epoch 871/1000
Epoch 872/1000
```

```
Epoch 873/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.5685e-04
Epoch 874/1000
Epoch 875/1000
Epoch 876/1000
Epoch 877/1000
Epoch 878/1000
Epoch 879/1000
Epoch 880/1000
Epoch 881/1000
Epoch 882/1000
83/83 [============== ] - 10s 117ms/step - loss: 2.8101e-04
Epoch 883/1000
Epoch 884/1000
83/83 [============= ] - 10s 117ms/step - loss: 2.8045e-04
Epoch 885/1000
Epoch 886/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.7014e-04
Epoch 887/1000
Epoch 888/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.7483e-04
Epoch 889/1000
Epoch 890/1000
Epoch 891/1000
Epoch 892/1000
Epoch 893/1000
Epoch 894/1000
Epoch 895/1000
Epoch 896/1000
```

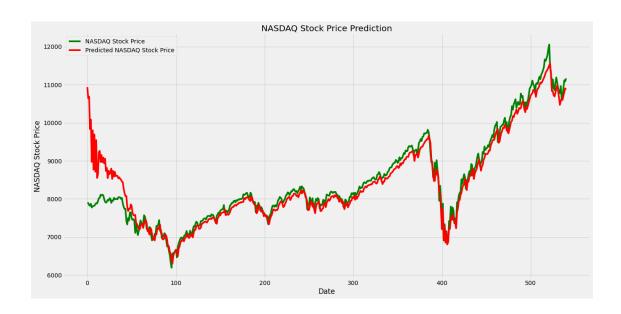
```
Epoch 897/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.7537e-04
Epoch 898/1000
Epoch 899/1000
Epoch 900/1000
Epoch 901/1000
Epoch 902/1000
Epoch 903/1000
Epoch 904/1000
Epoch 905/1000
Epoch 906/1000
83/83 [============== ] - 10s 121ms/step - loss: 2.3482e-04
Epoch 907/1000
Epoch 908/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.7610e-04
Epoch 909/1000
Epoch 910/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.7422e-04
Epoch 911/1000
Epoch 912/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.6189e-04
Epoch 913/1000
Epoch 914/1000
Epoch 915/1000
Epoch 916/1000
Epoch 917/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.8872e-04
Epoch 918/1000
Epoch 919/1000
Epoch 920/1000
```

```
Epoch 921/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.6865e-04
Epoch 922/1000
Epoch 923/1000
Epoch 924/1000
Epoch 925/1000
Epoch 926/1000
Epoch 927/1000
Epoch 928/1000
Epoch 929/1000
Epoch 930/1000
83/83 [============== ] - 10s 118ms/step - loss: 2.4670e-04
Epoch 931/1000
Epoch 932/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.6036e-04
Epoch 933/1000
Epoch 934/1000
83/83 [============ ] - 10s 122ms/step - loss: 2.5750e-04
Epoch 935/1000
Epoch 936/1000
Epoch 937/1000
Epoch 938/1000
Epoch 939/1000
Epoch 940/1000
Epoch 941/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.7916e-04
Epoch 942/1000
Epoch 943/1000
Epoch 944/1000
```

```
Epoch 945/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.6430e-04
Epoch 946/1000
Epoch 947/1000
Epoch 948/1000
Epoch 949/1000
Epoch 950/1000
Epoch 951/1000
Epoch 952/1000
Epoch 953/1000
Epoch 954/1000
83/83 [============== ] - 10s 118ms/step - loss: 2.5776e-04
Epoch 955/1000
Epoch 956/1000
83/83 [============= ] - 10s 118ms/step - loss: 2.7351e-04
Epoch 957/1000
Epoch 958/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.5296e-04
Epoch 959/1000
Epoch 960/1000
Epoch 961/1000
Epoch 962/1000
Epoch 963/1000
Epoch 964/1000
Epoch 965/1000
Epoch 966/1000
Epoch 967/1000
Epoch 968/1000
```

```
Epoch 969/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.8120e-04
Epoch 970/1000
Epoch 971/1000
Epoch 972/1000
Epoch 973/1000
Epoch 974/1000
Epoch 975/1000
Epoch 976/1000
Epoch 977/1000
Epoch 978/1000
83/83 [============== ] - 10s 118ms/step - loss: 2.4874e-04
Epoch 979/1000
Epoch 980/1000
83/83 [============ ] - 10s 117ms/step - loss: 2.4184e-04
Epoch 981/1000
Epoch 982/1000
83/83 [============ ] - 10s 118ms/step - loss: 2.7406e-04
Epoch 983/1000
Epoch 984/1000
Epoch 985/1000
Epoch 986/1000
Epoch 987/1000
Epoch 988/1000
Epoch 989/1000
Epoch 990/1000
Epoch 991/1000
Epoch 992/1000
```

```
Epoch 993/1000
    Epoch 994/1000
    Epoch 995/1000
    Epoch 996/1000
    Epoch 997/1000
    83/83 [============ ] - 10s 120ms/step - loss: 2.6350e-04
    Epoch 998/1000
    Epoch 999/1000
    83/83 [============ ] - 10s 117ms/step - loss: 2.5074e-04
    Epoch 1000/1000
    [92]: <tensorflow.python.keras.callbacks.History at 0x7fdeb1bae190>
[93]: testdataframe= test
    testdataframe['Date'] = testdataframe.index
    testdata = pd.DataFrame(columns = ['Date', 'Close'])
    testdata['Date'] = testdataframe['Date']
    testdata['Close'] = testdataframe['Close']
    real_stock_price = testdata.iloc[:, 1:2].values
    dataset_total = pd.concat((data2['Close'], testdata['Close']), axis = 0)
    inputs = dataset_total[len(dataset_total) - len(testdata) - 60:].values
    inputs = inputs.reshape(-1,1)
    inputs = sc.transform(inputs)
    X_{test} = []
    for i in range(60, inputs.shape[0]):
       X_test.append(inputs[i-60:i, 0])
    X test = np.array(X test)
    X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], 1))
[94]: predicted_with_gru = regressorGRU.predict(X_test)
    predicted_with_gru = sc.inverse_transform(predicted_with_gru)
[95]: plt.figure(figsize=(20,10))
    plt.plot(real_stock_price, color = 'green', label = 'NASDAQ Stock Price')
    plt.plot(predicted_with_gru, color = 'red', label = 'Predicted NASDAQ Stock_
     →Price')
    plt.title('NASDAQ Stock Price Prediction')
    plt.xlabel('Date')
    plt.ylabel('NASDAQ Stock Price')
    plt.legend()
    plt.show()
```



```
[96]: rmse_predict1= np.reshape(predicted_with_gru,541)

[97]: #forecast metrics
    def smape(a, f):
        return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)
        smape(test["Close"].values,rmse_predict1)

[97]: 2.4953968614719706

[98]: #forecast metrics
    from sklearn.metrics import mean_absolute_error
    from sklearn.metrics import mean_squared_error
    from sklearn.metrics import r2_score
    mae = mean_absolute_error(test["Close"].values, rmse_predict1)
    mse = mean_squared_error(test["Close"].values, rmse_predict1)
    rmse = np.sqrt(mse)
```

Results of sklearn.metrics: MAE: 214.1805178292514

print("Results of sklearn.metrics:")

MSE: 148283.06866175527 RMSE: 385.07540646184515

print("MAE:",mae)
print("MSE:", mse)
print("RMSE:", rmse)

```
[99]: elapsed_time = time.process_time() - t
       print(elapsed_time)
      31677.686889071
  []:
  []:
[100]: #repeat the same process for TSX
       df=pd.read_csv("GSPTSE_v1.csv", sep=",")
       from datetime import datetime
       con=df['Date']
       df['Date'] = pd.to datetime(df['Date'])
       df.set_index('Date', inplace=True)
       test = df[2164:]
       train = df[:2163]
[101]: df['Date'] = df.index
       data2 = pd.DataFrame(columns = ['Date', 'Close'])
       data2['Date'] = df['Date']
       data2['Close'] = df['Close']
[102]: import time
       t = time.process_time()
[103]: train_set = data2.iloc[:, 1:2].values
       sc = MinMaxScaler(feature range = (0, 1))
       training_set_scaled = sc.fit_transform(train_set)
       X_train = []
       y_train = []
       for i in range(60, training_set_scaled.shape[0]):
           X_train.append(training_set_scaled[i-60:i, 0])
           y_train.append(training_set_scaled[i, 0])
       X_train, y_train = np.array(X_train), np.array(y_train)
       X_train = np.reshape(X_train, (X_train.shape[0], X_train.shape[1], 1))
[104]: regressor = Sequential()
       regressor.add(LSTM(units = 50, return_sequences = True, input_shape = (X_train.
        \rightarrowshape[1], 1)))
       regressor.add(Dropout(0.15))
       regressor.add(LSTM(units = 50, return_sequences = True))
       regressor.add(Dropout(0.15))
       regressor.add(LSTM(units = 50, return_sequences = True))
       regressor.add(Dropout(0.15))
       regressor.add(LSTM(units = 50))
```

```
regressor.add(Dropout(0.15))
  regressor.add(Dense(units = 1))
[105]: regressor.compile(optimizer = 'adam', loss = 'mean_squared_error')
  regressor.fit(X_train, y_train, epochs = 1000, batch_size = 32)
  Epoch 1/1000
  83/83 [============ ] - 11s 134ms/step - loss: 0.0201
  Epoch 2/1000
  83/83 [============== ] - 11s 129ms/step - loss: 0.0061
  Epoch 3/1000
  Epoch 4/1000
  Epoch 5/1000
  Epoch 6/1000
  Epoch 7/1000
  Epoch 8/1000
  Epoch 9/1000
  83/83 [============== ] - 10s 124ms/step - loss: 0.0034
  Epoch 10/1000
  Epoch 11/1000
  Epoch 12/1000
  Epoch 13/1000
  Epoch 14/1000
  83/83 [=========== ] - 10s 126ms/step - loss: 0.0027
  Epoch 15/1000
  83/83 [============== ] - 10s 124ms/step - loss: 0.0024
  Epoch 16/1000
  Epoch 17/1000
  Epoch 18/1000
  83/83 [============== ] - 10s 124ms/step - loss: 0.0025
  Epoch 19/1000
  Epoch 20/1000
  Epoch 21/1000
```

```
Epoch 22/1000
83/83 [=========== ] - 10s 122ms/step - loss: 0.0021
Epoch 23/1000
83/83 [============== ] - 10s 123ms/step - loss: 0.0019
Epoch 24/1000
Epoch 25/1000
83/83 [============== ] - 11s 130ms/step - loss: 0.0018
Epoch 26/1000
83/83 [============ ] - 11s 134ms/step - loss: 0.0018
Epoch 27/1000
83/83 [============= ] - 10s 123ms/step - loss: 0.0019
Epoch 28/1000
Epoch 29/1000
Epoch 30/1000
Epoch 31/1000
Epoch 32/1000
Epoch 33/1000
83/83 [============= ] - 10s 121ms/step - loss: 0.0015
Epoch 34/1000
83/83 [============= ] - 10s 121ms/step - loss: 0.0016
Epoch 35/1000
Epoch 36/1000
Epoch 37/1000
Epoch 38/1000
83/83 [============== ] - 10s 121ms/step - loss: 0.0013
Epoch 39/1000
83/83 [============== ] - 10s 121ms/step - loss: 0.0014
Epoch 40/1000
Epoch 41/1000
Epoch 42/1000
Epoch 43/1000
Epoch 44/1000
Epoch 45/1000
```

```
Epoch 46/1000
83/83 [============ ] - 10s 121ms/step - loss: 0.0013
Epoch 47/1000
83/83 [============== ] - 10s 121ms/step - loss: 0.0011
Epoch 48/1000
Epoch 49/1000
Epoch 50/1000
83/83 [============ ] - 10s 121ms/step - loss: 0.0011
Epoch 51/1000
83/83 [============== ] - 10s 121ms/step - loss: 0.0011
Epoch 52/1000
Epoch 53/1000
83/83 [============== ] - 10s 121ms/step - loss: 0.0011
Epoch 54/1000
Epoch 55/1000
Epoch 56/1000
Epoch 57/1000
83/83 [============= ] - 10s 121ms/step - loss: 0.0011
Epoch 58/1000
Epoch 59/1000
Epoch 60/1000
Epoch 61/1000
Epoch 62/1000
Epoch 63/1000
Epoch 64/1000
Epoch 65/1000
Epoch 66/1000
Epoch 67/1000
83/83 [============ ] - 10s 119ms/step - loss: 8.9804e-04
Epoch 68/1000
Epoch 69/1000
```

```
Epoch 70/1000
83/83 [============ ] - 10s 120ms/step - loss: 9.0170e-04
Epoch 71/1000
Epoch 72/1000
Epoch 73/1000
Epoch 74/1000
83/83 [============ ] - 10s 119ms/step - loss: 8.5034e-04
Epoch 75/1000
Epoch 76/1000
Epoch 77/1000
83/83 [============ ] - 10s 120ms/step - loss: 8.9617e-04
Epoch 78/1000
Epoch 79/1000
Epoch 80/1000
Epoch 81/1000
Epoch 82/1000
Epoch 83/1000
Epoch 84/1000
Epoch 85/1000
Epoch 86/1000
Epoch 87/1000
Epoch 88/1000
Epoch 89/1000
Epoch 90/1000
Epoch 91/1000
83/83 [============ ] - 10s 118ms/step - loss: 7.9828e-04
Epoch 92/1000
Epoch 93/1000
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Epoch 94/1000
83/83 [============ ] - 10s 118ms/step - loss: 8.1061e-04
Epoch 95/1000
83/83 [============== ] - 10s 118ms/step - loss: 8.6319e-04
Epoch 96/1000
Epoch 97/1000
Epoch 98/1000
Epoch 99/1000
Epoch 100/1000
Epoch 101/1000
83/83 [============ ] - 10s 119ms/step - loss: 7.1892e-04
Epoch 102/1000
Epoch 103/1000
Epoch 104/1000
Epoch 105/1000
Epoch 106/1000
Epoch 107/1000
Epoch 108/1000
Epoch 109/1000
Epoch 110/1000
83/83 [============= ] - 10s 118ms/step - loss: 7.7854e-04
Epoch 111/1000
Epoch 112/1000
Epoch 113/1000
Epoch 114/1000
Epoch 115/1000
83/83 [============ ] - 10s 118ms/step - loss: 7.1497e-04
Epoch 116/1000
Epoch 117/1000
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Epoch 118/1000
83/83 [============ ] - 10s 118ms/step - loss: 7.6271e-04
Epoch 119/1000
Epoch 120/1000
Epoch 121/1000
Epoch 122/1000
Epoch 123/1000
83/83 [============= ] - 10s 118ms/step - loss: 7.9714e-04
Epoch 124/1000
Epoch 125/1000
83/83 [============ ] - 10s 118ms/step - loss: 7.7505e-04
Epoch 126/1000
Epoch 127/1000
Epoch 128/1000
Epoch 129/1000
Epoch 130/1000
Epoch 131/1000
Epoch 132/1000
Epoch 133/1000
Epoch 134/1000
83/83 [============= ] - 10s 118ms/step - loss: 7.1614e-04
Epoch 135/1000
Epoch 136/1000
Epoch 137/1000
Epoch 138/1000
Epoch 139/1000
83/83 [============ ] - 10s 117ms/step - loss: 7.2662e-04
Epoch 140/1000
Epoch 141/1000
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Epoch 142/1000
Epoch 143/1000
Epoch 144/1000
Epoch 145/1000
Epoch 146/1000
Epoch 147/1000
Epoch 148/1000
Epoch 149/1000
83/83 [============ ] - 10s 118ms/step - loss: 7.3495e-04
Epoch 150/1000
Epoch 151/1000
Epoch 152/1000
Epoch 153/1000
Epoch 154/1000
Epoch 155/1000
Epoch 156/1000
Epoch 157/1000
Epoch 158/1000
83/83 [============= ] - 10s 119ms/step - loss: 7.2222e-04
Epoch 159/1000
Epoch 160/1000
Epoch 161/1000
Epoch 162/1000
Epoch 163/1000
83/83 [============ ] - 10s 118ms/step - loss: 6.8476e-04
Epoch 164/1000
Epoch 165/1000
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Epoch 166/1000
Epoch 167/1000
Epoch 168/1000
83/83 [============== ] - 10s 117ms/step - loss: 6.7756e-04
Epoch 169/1000
Epoch 170/1000
Epoch 171/1000
Epoch 172/1000
Epoch 173/1000
83/83 [============= ] - 10s 126ms/step - loss: 7.2563e-04
Epoch 174/1000
Epoch 175/1000
Epoch 176/1000
Epoch 177/1000
Epoch 178/1000
Epoch 179/1000
Epoch 180/1000
Epoch 181/1000
Epoch 182/1000
Epoch 183/1000
Epoch 184/1000
Epoch 185/1000
Epoch 186/1000
Epoch 187/1000
83/83 [============ ] - 10s 117ms/step - loss: 6.4451e-04
Epoch 188/1000
Epoch 189/1000
```

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Epoch 190/1000
83/83 [============ ] - 10s 118ms/step - loss: 6.2219e-04
Epoch 191/1000
Epoch 192/1000
83/83 [============== ] - 10s 117ms/step - loss: 6.4951e-04
Epoch 193/1000
Epoch 194/1000
Epoch 195/1000
Epoch 196/1000
Epoch 197/1000
83/83 [============= ] - 10s 120ms/step - loss: 6.5702e-04
Epoch 198/1000
Epoch 199/1000
Epoch 200/1000
Epoch 201/1000
Epoch 202/1000
Epoch 203/1000
Epoch 204/1000
Epoch 205/1000
Epoch 206/1000
83/83 [============= ] - 10s 118ms/step - loss: 6.8496e-04
Epoch 207/1000
Epoch 208/1000
Epoch 209/1000
Epoch 210/1000
Epoch 211/1000
83/83 [============ ] - 10s 117ms/step - loss: 6.8482e-04
Epoch 212/1000
Epoch 213/1000
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Epoch 214/1000
Epoch 215/1000
Epoch 216/1000
Epoch 217/1000
Epoch 218/1000
Epoch 219/1000
Epoch 220/1000
Epoch 221/1000
83/83 [============ ] - 10s 118ms/step - loss: 6.6231e-04
Epoch 222/1000
Epoch 223/1000
Epoch 224/1000
Epoch 225/1000
Epoch 226/1000
Epoch 227/1000
Epoch 228/1000
Epoch 229/1000
Epoch 230/1000
83/83 [============= ] - 10s 119ms/step - loss: 6.5054e-04
Epoch 231/1000
Epoch 232/1000
Epoch 233/1000
Epoch 234/1000
Epoch 235/1000
83/83 [============ ] - 10s 124ms/step - loss: 6.6377e-04
Epoch 236/1000
Epoch 237/1000
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Epoch 238/1000
Epoch 239/1000
Epoch 240/1000
83/83 [============== ] - 11s 137ms/step - loss: 6.1287e-04
Epoch 241/1000
Epoch 242/1000
Epoch 243/1000
Epoch 244/1000
Epoch 245/1000
83/83 [============ ] - 10s 118ms/step - loss: 6.2232e-04
Epoch 246/1000
Epoch 247/1000
Epoch 248/1000
Epoch 249/1000
Epoch 250/1000
Epoch 251/1000
Epoch 252/1000
Epoch 253/1000
Epoch 254/1000
83/83 [============== ] - 10s 119ms/step - loss: 6.3647e-04
Epoch 255/1000
Epoch 256/1000
Epoch 257/1000
Epoch 258/1000
Epoch 259/1000
83/83 [============ ] - 10s 117ms/step - loss: 6.5490e-04
Epoch 260/1000
Epoch 261/1000
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Epoch 262/1000
Epoch 263/1000
Epoch 264/1000
Epoch 265/1000
Epoch 266/1000
Epoch 267/1000
Epoch 268/1000
Epoch 269/1000
83/83 [============ ] - 11s 127ms/step - loss: 6.1358e-04
Epoch 270/1000
Epoch 271/1000
Epoch 272/1000
Epoch 273/1000
Epoch 274/1000
Epoch 275/1000
Epoch 276/1000
Epoch 277/1000
Epoch 278/1000
83/83 [============== ] - 10s 116ms/step - loss: 5.9598e-04
Epoch 279/1000
Epoch 280/1000
Epoch 281/1000
Epoch 282/1000
Epoch 283/1000
83/83 [============ ] - 10s 116ms/step - loss: 5.7948e-04
Epoch 284/1000
Epoch 285/1000
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Epoch 286/1000
Epoch 287/1000
Epoch 288/1000
83/83 [============== ] - 10s 117ms/step - loss: 6.3098e-04
Epoch 289/1000
Epoch 290/1000
Epoch 291/1000
Epoch 292/1000
Epoch 293/1000
83/83 [============ ] - 10s 117ms/step - loss: 6.0518e-04
Epoch 294/1000
Epoch 295/1000
Epoch 296/1000
Epoch 297/1000
Epoch 298/1000
Epoch 299/1000
Epoch 300/1000
Epoch 301/1000
Epoch 302/1000
Epoch 303/1000
83/83 [============== ] - 11s 131ms/step - loss: 6.2110e-04
Epoch 304/1000
Epoch 305/1000
Epoch 306/1000
Epoch 307/1000
83/83 [============ ] - 10s 116ms/step - loss: 6.1375e-04
Epoch 308/1000
Epoch 309/1000
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Epoch 310/1000
Epoch 311/1000
Epoch 312/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.9937e-04
Epoch 313/1000
Epoch 314/1000
Epoch 315/1000
Epoch 316/1000
Epoch 317/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.6810e-04
Epoch 318/1000
Epoch 319/1000
Epoch 320/1000
Epoch 321/1000
Epoch 322/1000
Epoch 323/1000
Epoch 324/1000
Epoch 325/1000
Epoch 326/1000
83/83 [============== ] - 10s 117ms/step - loss: 6.3241e-04
Epoch 327/1000
Epoch 328/1000
Epoch 329/1000
Epoch 330/1000
Epoch 331/1000
83/83 [============ ] - 11s 127ms/step - loss: 5.9784e-04
Epoch 332/1000
Epoch 333/1000
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Epoch 334/1000
Epoch 335/1000
Epoch 336/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.9384e-04
Epoch 337/1000
Epoch 338/1000
Epoch 339/1000
Epoch 340/1000
Epoch 341/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.7731e-04
Epoch 342/1000
Epoch 343/1000
Epoch 344/1000
Epoch 345/1000
Epoch 346/1000
Epoch 347/1000
Epoch 348/1000
Epoch 349/1000
Epoch 350/1000
83/83 [============== ] - 10s 117ms/step - loss: 6.1488e-04
Epoch 351/1000
Epoch 352/1000
Epoch 353/1000
Epoch 354/1000
Epoch 355/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.9969e-04
Epoch 356/1000
Epoch 357/1000
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Epoch 358/1000
Epoch 359/1000
Epoch 360/1000
Epoch 361/1000
Epoch 362/1000
Epoch 363/1000
Epoch 364/1000
Epoch 365/1000
83/83 [============ ] - 12s 139ms/step - loss: 6.1942e-04
Epoch 366/1000
Epoch 367/1000
Epoch 368/1000
Epoch 369/1000
Epoch 370/1000
Epoch 371/1000
Epoch 372/1000
Epoch 373/1000
Epoch 374/1000
83/83 [============== ] - 10s 118ms/step - loss: 6.1878e-04
Epoch 375/1000
Epoch 376/1000
Epoch 377/1000
Epoch 378/1000
Epoch 379/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.6523e-04
Epoch 380/1000
Epoch 381/1000
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Epoch 382/1000
Epoch 383/1000
Epoch 384/1000
83/83 [============= ] - 10s 118ms/step - loss: 5.9224e-04
Epoch 385/1000
Epoch 386/1000
Epoch 387/1000
Epoch 388/1000
Epoch 389/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.3709e-04
Epoch 390/1000
Epoch 391/1000
Epoch 392/1000
Epoch 393/1000
Epoch 394/1000
Epoch 395/1000
Epoch 396/1000
Epoch 397/1000
Epoch 398/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.4200e-04
Epoch 399/1000
Epoch 400/1000
Epoch 401/1000
Epoch 402/1000
Epoch 403/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.4997e-04
Epoch 404/1000
Epoch 405/1000
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Epoch 406/1000
Epoch 407/1000
Epoch 408/1000
Epoch 409/1000
Epoch 410/1000
Epoch 411/1000
Epoch 412/1000
Epoch 413/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.4619e-04
Epoch 414/1000
Epoch 415/1000
Epoch 416/1000
Epoch 417/1000
Epoch 418/1000
Epoch 419/1000
Epoch 420/1000
Epoch 421/1000
Epoch 422/1000
Epoch 423/1000
Epoch 424/1000
Epoch 425/1000
Epoch 426/1000
Epoch 427/1000
83/83 [============ ] - 11s 137ms/step - loss: 5.7165e-04
Epoch 428/1000
Epoch 429/1000
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Epoch 430/1000
Epoch 431/1000
Epoch 432/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.7065e-04
Epoch 433/1000
Epoch 434/1000
Epoch 435/1000
Epoch 436/1000
Epoch 437/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.3487e-04
Epoch 438/1000
Epoch 439/1000
Epoch 440/1000
Epoch 441/1000
Epoch 442/1000
Epoch 443/1000
Epoch 444/1000
Epoch 445/1000
Epoch 446/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.6142e-04
Epoch 447/1000
Epoch 448/1000
Epoch 449/1000
Epoch 450/1000
Epoch 451/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.3933e-04
Epoch 452/1000
Epoch 453/1000
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Epoch 454/1000
Epoch 455/1000
Epoch 456/1000
83/83 [============= ] - 10s 118ms/step - loss: 5.7064e-04
Epoch 457/1000
Epoch 458/1000
Epoch 459/1000
Epoch 460/1000
Epoch 461/1000
83/83 [============ ] - 10s 120ms/step - loss: 5.1019e-04
Epoch 462/1000
Epoch 463/1000
Epoch 464/1000
Epoch 465/1000
Epoch 466/1000
Epoch 467/1000
Epoch 468/1000
Epoch 469/1000
Epoch 470/1000
83/83 [============= ] - 10s 118ms/step - loss: 5.5500e-04
Epoch 471/1000
Epoch 472/1000
Epoch 473/1000
Epoch 474/1000
Epoch 475/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.3449e-04
Epoch 476/1000
Epoch 477/1000
```

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Epoch 478/1000
Epoch 479/1000
Epoch 480/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.3843e-04
Epoch 481/1000
Epoch 482/1000
Epoch 483/1000
Epoch 484/1000
Epoch 485/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.4809e-04
Epoch 486/1000
Epoch 487/1000
Epoch 488/1000
Epoch 489/1000
Epoch 490/1000
Epoch 491/1000
Epoch 492/1000
Epoch 493/1000
Epoch 494/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.0197e-04
Epoch 495/1000
Epoch 496/1000
Epoch 497/1000
Epoch 498/1000
Epoch 499/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.6135e-04
Epoch 500/1000
Epoch 501/1000
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Epoch 502/1000
Epoch 503/1000
Epoch 504/1000
Epoch 505/1000
Epoch 506/1000
Epoch 507/1000
Epoch 508/1000
Epoch 509/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.2384e-04
Epoch 510/1000
Epoch 511/1000
Epoch 512/1000
Epoch 513/1000
Epoch 514/1000
Epoch 515/1000
Epoch 516/1000
Epoch 517/1000
Epoch 518/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.1908e-04
Epoch 519/1000
Epoch 520/1000
Epoch 521/1000
Epoch 522/1000
Epoch 523/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.4325e-04
Epoch 524/1000
Epoch 525/1000
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Epoch 526/1000
Epoch 527/1000
Epoch 528/1000
Epoch 529/1000
Epoch 530/1000
Epoch 531/1000
Epoch 532/1000
Epoch 533/1000
83/83 [============ ] - 10s 118ms/step - loss: 4.8211e-04
Epoch 534/1000
Epoch 535/1000
Epoch 536/1000
Epoch 537/1000
Epoch 538/1000
Epoch 539/1000
Epoch 540/1000
Epoch 541/1000
Epoch 542/1000
83/83 [============== ] - 10s 120ms/step - loss: 5.1521e-04
Epoch 543/1000
Epoch 544/1000
Epoch 545/1000
Epoch 546/1000
Epoch 547/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.5054e-04
Epoch 548/1000
Epoch 549/1000
```

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Epoch 550/1000
Epoch 551/1000
Epoch 552/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.1292e-04
Epoch 553/1000
Epoch 554/1000
Epoch 555/1000
Epoch 556/1000
Epoch 557/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.4711e-04
Epoch 558/1000
Epoch 559/1000
Epoch 560/1000
Epoch 561/1000
Epoch 562/1000
Epoch 563/1000
Epoch 564/1000
Epoch 565/1000
Epoch 566/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.1923e-04
Epoch 567/1000
Epoch 568/1000
Epoch 569/1000
Epoch 570/1000
Epoch 571/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.4576e-04
Epoch 572/1000
Epoch 573/1000
```

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Epoch 574/1000
Epoch 575/1000
Epoch 576/1000
83/83 [============= ] - 10s 118ms/step - loss: 5.3596e-04
Epoch 577/1000
Epoch 578/1000
Epoch 579/1000
Epoch 580/1000
Epoch 581/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.6321e-04
Epoch 582/1000
Epoch 583/1000
Epoch 584/1000
Epoch 585/1000
Epoch 586/1000
Epoch 587/1000
Epoch 588/1000
Epoch 589/1000
Epoch 590/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.2442e-04
Epoch 591/1000
Epoch 592/1000
Epoch 593/1000
Epoch 594/1000
Epoch 595/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.0721e-04
Epoch 596/1000
Epoch 597/1000
```

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Epoch 598/1000
83/83 [============ ] - 10s 120ms/step - loss: 4.8201e-04
Epoch 599/1000
Epoch 600/1000
Epoch 601/1000
Epoch 602/1000
Epoch 603/1000
Epoch 604/1000
Epoch 605/1000
83/83 [============ ] - 10s 120ms/step - loss: 5.2822e-04
Epoch 606/1000
Epoch 607/1000
Epoch 608/1000
Epoch 609/1000
Epoch 610/1000
Epoch 611/1000
Epoch 612/1000
Epoch 613/1000
Epoch 614/1000
Epoch 615/1000
Epoch 616/1000
Epoch 617/1000
Epoch 618/1000
Epoch 619/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8427e-04
Epoch 620/1000
Epoch 621/1000
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Epoch 622/1000
Epoch 623/1000
Epoch 624/1000
Epoch 625/1000
Epoch 626/1000
Epoch 627/1000
Epoch 628/1000
Epoch 629/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.1122e-04
Epoch 630/1000
Epoch 631/1000
Epoch 632/1000
Epoch 633/1000
Epoch 634/1000
Epoch 635/1000
Epoch 636/1000
Epoch 637/1000
Epoch 638/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.3648e-04
Epoch 639/1000
Epoch 640/1000
Epoch 641/1000
Epoch 642/1000
Epoch 643/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.2041e-04
Epoch 644/1000
Epoch 645/1000
```

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Epoch 646/1000
Epoch 647/1000
Epoch 648/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.0545e-04
Epoch 649/1000
Epoch 650/1000
Epoch 651/1000
Epoch 652/1000
Epoch 653/1000
83/83 [============ ] - 10s 116ms/step - loss: 5.1415e-04
Epoch 654/1000
Epoch 655/1000
Epoch 656/1000
Epoch 657/1000
Epoch 658/1000
Epoch 659/1000
Epoch 660/1000
Epoch 661/1000
Epoch 662/1000
83/83 [============= ] - 10s 118ms/step - loss: 5.3424e-04
Epoch 663/1000
Epoch 664/1000
Epoch 665/1000
Epoch 666/1000
Epoch 667/1000
83/83 [============ ] - 11s 127ms/step - loss: 5.3309e-04
Epoch 668/1000
Epoch 669/1000
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Epoch 670/1000
Epoch 671/1000
Epoch 672/1000
Epoch 673/1000
Epoch 674/1000
Epoch 675/1000
Epoch 676/1000
Epoch 677/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.2269e-04
Epoch 678/1000
Epoch 679/1000
Epoch 680/1000
Epoch 681/1000
Epoch 682/1000
Epoch 683/1000
Epoch 684/1000
Epoch 685/1000
Epoch 686/1000
83/83 [============== ] - 10s 119ms/step - loss: 5.8974e-04
Epoch 687/1000
Epoch 688/1000
Epoch 689/1000
Epoch 690/1000
Epoch 691/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.0769e-04
Epoch 692/1000
Epoch 693/1000
```

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Epoch 694/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.1683e-04
Epoch 695/1000
Epoch 696/1000
83/83 [============== ] - 10s 116ms/step - loss: 4.8562e-04
Epoch 697/1000
Epoch 698/1000
Epoch 699/1000
Epoch 700/1000
Epoch 701/1000
83/83 [============= ] - 10s 117ms/step - loss: 5.1427e-04
Epoch 702/1000
Epoch 703/1000
Epoch 704/1000
Epoch 705/1000
Epoch 706/1000
Epoch 707/1000
Epoch 708/1000
Epoch 709/1000
Epoch 710/1000
Epoch 711/1000
Epoch 712/1000
Epoch 713/1000
Epoch 714/1000
Epoch 715/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.2414e-04
Epoch 716/1000
Epoch 717/1000
```

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Epoch 718/1000
Epoch 719/1000
Epoch 720/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.6876e-04
Epoch 721/1000
Epoch 722/1000
Epoch 723/1000
Epoch 724/1000
Epoch 725/1000
83/83 [=========== ] - 10s 118ms/step - loss: 4.6900e-04
Epoch 726/1000
Epoch 727/1000
Epoch 728/1000
Epoch 729/1000
Epoch 730/1000
Epoch 731/1000
Epoch 732/1000
Epoch 733/1000
Epoch 734/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.0358e-04
Epoch 735/1000
Epoch 736/1000
Epoch 737/1000
Epoch 738/1000
Epoch 739/1000
83/83 [============ ] - 10s 118ms/step - loss: 4.9886e-04
Epoch 740/1000
Epoch 741/1000
```

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Epoch 742/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8185e-04
Epoch 743/1000
Epoch 744/1000
Epoch 745/1000
Epoch 746/1000
Epoch 747/1000
Epoch 748/1000
Epoch 749/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.0004e-04
Epoch 750/1000
Epoch 751/1000
Epoch 752/1000
Epoch 753/1000
Epoch 754/1000
Epoch 755/1000
Epoch 756/1000
Epoch 757/1000
Epoch 758/1000
Epoch 759/1000
Epoch 760/1000
Epoch 761/1000
Epoch 762/1000
Epoch 763/1000
83/83 [============ ] - 11s 127ms/step - loss: 5.0990e-04
Epoch 764/1000
Epoch 765/1000
```

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Epoch 766/1000
83/83 [============ ] - 10s 123ms/step - loss: 5.2322e-04
Epoch 767/1000
Epoch 768/1000
Epoch 769/1000
Epoch 770/1000
Epoch 771/1000
Epoch 772/1000
Epoch 773/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.9697e-04
Epoch 774/1000
Epoch 775/1000
Epoch 776/1000
Epoch 777/1000
Epoch 778/1000
Epoch 779/1000
Epoch 780/1000
Epoch 781/1000
Epoch 782/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.8926e-04
Epoch 783/1000
Epoch 784/1000
Epoch 785/1000
Epoch 786/1000
Epoch 787/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.5510e-04
Epoch 788/1000
Epoch 789/1000
```

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Epoch 790/1000
Epoch 791/1000
Epoch 792/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.1761e-04
Epoch 793/1000
Epoch 794/1000
Epoch 795/1000
Epoch 796/1000
Epoch 797/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.9553e-04
Epoch 798/1000
Epoch 799/1000
Epoch 800/1000
Epoch 801/1000
Epoch 802/1000
Epoch 803/1000
Epoch 804/1000
Epoch 805/1000
Epoch 806/1000
83/83 [============= ] - 10s 117ms/step - loss: 4.8600e-04
Epoch 807/1000
Epoch 808/1000
Epoch 809/1000
Epoch 810/1000
Epoch 811/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8840e-04
Epoch 812/1000
Epoch 813/1000
```

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Epoch 814/1000
Epoch 815/1000
Epoch 816/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.1030e-04
Epoch 817/1000
Epoch 818/1000
Epoch 819/1000
Epoch 820/1000
Epoch 821/1000
83/83 [============ ] - 10s 116ms/step - loss: 5.1834e-04
Epoch 822/1000
Epoch 823/1000
Epoch 824/1000
Epoch 825/1000
Epoch 826/1000
Epoch 827/1000
Epoch 828/1000
Epoch 829/1000
Epoch 830/1000
83/83 [============== ] - 11s 133ms/step - loss: 4.8798e-04
Epoch 831/1000
Epoch 832/1000
Epoch 833/1000
Epoch 834/1000
Epoch 835/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0430e-04
Epoch 836/1000
Epoch 837/1000
```

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Epoch 838/1000
Epoch 839/1000
Epoch 840/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.8720e-04
Epoch 841/1000
Epoch 842/1000
Epoch 843/1000
Epoch 844/1000
Epoch 845/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.7725e-04
Epoch 846/1000
Epoch 847/1000
Epoch 848/1000
Epoch 849/1000
Epoch 850/1000
Epoch 851/1000
Epoch 852/1000
Epoch 853/1000
Epoch 854/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.9735e-04
Epoch 855/1000
Epoch 856/1000
Epoch 857/1000
Epoch 858/1000
Epoch 859/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8344e-04
Epoch 860/1000
Epoch 861/1000
```

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Epoch 862/1000
Epoch 863/1000
Epoch 864/1000
83/83 [============== ] - 10s 116ms/step - loss: 4.8927e-04
Epoch 865/1000
Epoch 866/1000
Epoch 867/1000
Epoch 868/1000
Epoch 869/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0574e-04
Epoch 870/1000
Epoch 871/1000
Epoch 872/1000
Epoch 873/1000
Epoch 874/1000
Epoch 875/1000
Epoch 876/1000
Epoch 877/1000
Epoch 878/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.8053e-04
Epoch 879/1000
Epoch 880/1000
Epoch 881/1000
Epoch 882/1000
Epoch 883/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0712e-04
Epoch 884/1000
Epoch 885/1000
```

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Epoch 886/1000
Epoch 887/1000
Epoch 888/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.6682e-04
Epoch 889/1000
Epoch 890/1000
Epoch 891/1000
Epoch 892/1000
Epoch 893/1000
83/83 [============ ] - 11s 135ms/step - loss: 5.0713e-04
Epoch 894/1000
Epoch 895/1000
Epoch 896/1000
Epoch 897/1000
Epoch 898/1000
Epoch 899/1000
Epoch 900/1000
Epoch 901/1000
Epoch 902/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.7314e-04
Epoch 903/1000
Epoch 904/1000
Epoch 905/1000
Epoch 906/1000
Epoch 907/1000
83/83 [============ ] - 10s 116ms/step - loss: 5.0180e-04
Epoch 908/1000
Epoch 909/1000
```

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Epoch 910/1000
Epoch 911/1000
Epoch 912/1000
Epoch 913/1000
Epoch 914/1000
Epoch 915/1000
Epoch 916/1000
Epoch 917/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.8487e-04
Epoch 918/1000
Epoch 919/1000
Epoch 920/1000
Epoch 921/1000
Epoch 922/1000
Epoch 923/1000
Epoch 924/1000
Epoch 925/1000
Epoch 926/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.7424e-04
Epoch 927/1000
Epoch 928/1000
Epoch 929/1000
Epoch 930/1000
Epoch 931/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.1713e-04
Epoch 932/1000
Epoch 933/1000
```

```
Epoch 934/1000
Epoch 935/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.7827e-04
Epoch 936/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.9058e-04
Epoch 937/1000
Epoch 938/1000
Epoch 939/1000
Epoch 940/1000
Epoch 941/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.1393e-04
Epoch 942/1000
Epoch 943/1000
Epoch 944/1000
Epoch 945/1000
Epoch 946/1000
Epoch 947/1000
Epoch 948/1000
Epoch 949/1000
Epoch 950/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.6675e-04
Epoch 951/1000
Epoch 952/1000
Epoch 953/1000
Epoch 954/1000
Epoch 955/1000
83/83 [============ ] - 10s 120ms/step - loss: 4.9215e-04
Epoch 956/1000
Epoch 957/1000
```

```
Epoch 958/1000
Epoch 959/1000
83/83 [============== ] - 11s 127ms/step - loss: 5.0730e-04
Epoch 960/1000
83/83 [============== ] - 10s 126ms/step - loss: 4.7772e-04
Epoch 961/1000
Epoch 962/1000
Epoch 963/1000
Epoch 964/1000
Epoch 965/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.2595e-04
Epoch 966/1000
Epoch 967/1000
Epoch 968/1000
Epoch 969/1000
Epoch 970/1000
Epoch 971/1000
Epoch 972/1000
Epoch 973/1000
Epoch 974/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.7380e-04
Epoch 975/1000
Epoch 976/1000
Epoch 977/1000
Epoch 978/1000
Epoch 979/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8109e-04
Epoch 980/1000
Epoch 981/1000
```

```
Epoch 982/1000
 Epoch 983/1000
 Epoch 984/1000
 Epoch 985/1000
 Epoch 986/1000
 Epoch 987/1000
 Epoch 988/1000
 Epoch 989/1000
 83/83 [============ ] - 10s 117ms/step - loss: 5.0545e-04
 Epoch 990/1000
 Epoch 991/1000
 Epoch 992/1000
 Epoch 993/1000
 Epoch 994/1000
 Epoch 995/1000
 Epoch 996/1000
 Epoch 997/1000
 Epoch 998/1000
 Epoch 999/1000
 Epoch 1000/1000
 [105]: <tensorflow.python.keras.callbacks.History at 0x7fde97628310>
[106]: testdataframe= test
  testdataframe['Date'] = testdataframe.index
  testdata = pd.DataFrame(columns = ['Date', 'Close'])
  testdata['Date'] = testdataframe['Date']
  testdata['Close'] = testdataframe['Close']
```

```
real_stock_price = testdata.iloc[:, 1:2].values
dataset_total = pd.concat((data2['Close'], testdata['Close']), axis = 0)
inputs = dataset_total[len(dataset_total) - len(testdata) - 60:].values
inputs = inputs.reshape(-1,1)
inputs = sc.transform(inputs)
X_test = []
for i in range(60, 593):
    X_test.append(inputs[i-60:i, 0])
X_test = np.array(X_test)
X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], 1))
```

```
[107]: predicted_stock_price = regressor.predict(X_test)
predicted_stock_price = sc.inverse_transform(predicted_stock_price)
```



```
[109]: rmse_predict= np.reshape(predicted_stock_price,533)
[110]: #forecast metrics
def smape(a, f):
```

```
smape(test["Close"].values,rmse_predict)
[110]: 0.534842510006781
[111]: #forecast metrics
       from sklearn.metrics import mean_absolute_error
       from sklearn.metrics import mean_squared_error
       from sklearn.metrics import r2_score
       mae = mean_absolute_error(test["Close"].values, rmse predict)
       mse = mean_squared_error(test["Close"].values, rmse_predict)
       rmse = np.sqrt(mse)
       print("Results of sklearn.metrics:")
       print("MAE:",mae)
       print("MSE:", mse)
       print("RMSE:", rmse)
      Results of sklearn.metrics:
      MAE: 83.77830782070818
      MSE: 12129.751352569736
      RMSE: 110.13515039518371
[112]: elapsed time = time.process time() - t
       print(elapsed_time)
      30921.342013751993
  []:
  []:
[113]: import time
       t = time.process_time()
[114]: regressorGRU = Sequential()
       regressorGRU.add(GRU(units = 50, return_sequences = True, input_shape =__
       \hookrightarrow (X_train.shape[1],1)))
       regressorGRU.add(Dropout(0.15))
       regressorGRU.add(GRU(units = 50, return_sequences = True))
       regressorGRU.add(Dropout(0.15))
       regressorGRU.add(GRU(units = 50, return_sequences = True))
       regressorGRU.add(Dropout(0.15))
       regressorGRU.add(GRU(units = 50))
       regressorGRU.add(Dropout(0.15))
```

return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)

```
regressorGRU.add(Dense(units = 1))
[115]: regressorGRU.compile(optimizer='adam',loss='mean_squared_error')
[116]: regressorGRU.fit(X_train, y_train, epochs = 1000, batch_size = 32)
  Epoch 1/1000
  83/83 [============= ] - 13s 154ms/step - loss: 0.0207
  Epoch 2/1000
  Epoch 3/1000
  Epoch 4/1000
  Epoch 5/1000
  Epoch 6/1000
  83/83 [============ ] - 11s 138ms/step - loss: 0.0025
  Epoch 7/1000
  Epoch 8/1000
  Epoch 9/1000
  Epoch 10/1000
  Epoch 11/1000
  83/83 [============== ] - 11s 131ms/step - loss: 0.0021
  Epoch 12/1000
  Epoch 13/1000
  Epoch 14/1000
  Epoch 15/1000
  Epoch 16/1000
  Epoch 17/1000
  83/83 [============ ] - 11s 128ms/step - loss: 0.0017
  Epoch 18/1000
  Epoch 19/1000
  Epoch 20/1000
  Epoch 21/1000
```

```
Epoch 22/1000
83/83 [=========== ] - 11s 129ms/step - loss: 0.0015
Epoch 23/1000
Epoch 24/1000
Epoch 25/1000
83/83 [============== ] - 11s 129ms/step - loss: 0.0014
Epoch 26/1000
83/83 [============ ] - 11s 128ms/step - loss: 0.0013
Epoch 27/1000
Epoch 28/1000
Epoch 29/1000
Epoch 30/1000
Epoch 31/1000
Epoch 32/1000
Epoch 33/1000
83/83 [============ ] - 11s 127ms/step - loss: 0.0012
Epoch 34/1000
Epoch 35/1000
Epoch 36/1000
Epoch 37/1000
Epoch 38/1000
Epoch 39/1000
Epoch 40/1000
Epoch 41/1000
Epoch 42/1000
Epoch 43/1000
83/83 [============ ] - 11s 134ms/step - loss: 0.0012
Epoch 44/1000
83/83 [=============== ] - 11s 138ms/step - loss: 0.0011
Epoch 45/1000
```

```
Epoch 46/1000
83/83 [============ ] - 10s 126ms/step - loss: 9.9763e-04
Epoch 47/1000
Epoch 48/1000
Epoch 49/1000
Epoch 50/1000
83/83 [============ ] - 10s 125ms/step - loss: 0.0010
Epoch 51/1000
Epoch 52/1000
Epoch 53/1000
83/83 [============ ] - 10s 125ms/step - loss: 9.4649e-04
Epoch 54/1000
Epoch 55/1000
Epoch 56/1000
Epoch 57/1000
Epoch 58/1000
Epoch 59/1000
Epoch 60/1000
Epoch 61/1000
Epoch 62/1000
Epoch 63/1000
Epoch 64/1000
Epoch 65/1000
Epoch 66/1000
Epoch 67/1000
83/83 [============ ] - 10s 126ms/step - loss: 8.9825e-04
Epoch 68/1000
Epoch 69/1000
```

```
Epoch 70/1000
83/83 [============ ] - 11s 128ms/step - loss: 8.6691e-04
Epoch 71/1000
Epoch 72/1000
Epoch 73/1000
Epoch 74/1000
Epoch 75/1000
Epoch 76/1000
Epoch 77/1000
83/83 [============ ] - 10s 124ms/step - loss: 8.5753e-04
Epoch 78/1000
Epoch 79/1000
Epoch 80/1000
Epoch 81/1000
Epoch 82/1000
Epoch 83/1000
Epoch 84/1000
Epoch 85/1000
Epoch 86/1000
Epoch 87/1000
Epoch 88/1000
Epoch 89/1000
Epoch 90/1000
Epoch 91/1000
83/83 [============ ] - 10s 123ms/step - loss: 7.5131e-04
Epoch 92/1000
Epoch 93/1000
```

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Epoch 94/1000
83/83 [============ ] - 10s 124ms/step - loss: 7.6098e-04
Epoch 95/1000
Epoch 96/1000
Epoch 97/1000
Epoch 98/1000
Epoch 99/1000
Epoch 100/1000
Epoch 101/1000
83/83 [============= ] - 10s 120ms/step - loss: 7.6356e-04
Epoch 102/1000
Epoch 103/1000
Epoch 104/1000
Epoch 105/1000
Epoch 106/1000
Epoch 107/1000
Epoch 108/1000
Epoch 109/1000
Epoch 110/1000
Epoch 111/1000
Epoch 112/1000
Epoch 113/1000
Epoch 114/1000
Epoch 115/1000
83/83 [============ ] - 10s 121ms/step - loss: 7.7236e-04
Epoch 116/1000
Epoch 117/1000
```

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Epoch 118/1000
Epoch 119/1000
Epoch 120/1000
Epoch 121/1000
Epoch 122/1000
Epoch 123/1000
Epoch 124/1000
Epoch 125/1000
83/83 [============ ] - 10s 120ms/step - loss: 7.2522e-04
Epoch 126/1000
Epoch 127/1000
Epoch 128/1000
Epoch 129/1000
Epoch 130/1000
Epoch 131/1000
Epoch 132/1000
Epoch 133/1000
Epoch 134/1000
Epoch 135/1000
Epoch 136/1000
Epoch 137/1000
Epoch 138/1000
Epoch 139/1000
83/83 [============ ] - 10s 120ms/step - loss: 6.7644e-04
Epoch 140/1000
Epoch 141/1000
```

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Epoch 142/1000
Epoch 143/1000
Epoch 144/1000
Epoch 145/1000
Epoch 146/1000
Epoch 147/1000
Epoch 148/1000
Epoch 149/1000
83/83 [============ ] - 10s 119ms/step - loss: 6.6614e-04
Epoch 150/1000
Epoch 151/1000
Epoch 152/1000
Epoch 153/1000
Epoch 154/1000
Epoch 155/1000
Epoch 156/1000
Epoch 157/1000
Epoch 158/1000
Epoch 159/1000
Epoch 160/1000
Epoch 161/1000
Epoch 162/1000
Epoch 163/1000
83/83 [============ ] - 12s 143ms/step - loss: 6.8503e-04
Epoch 164/1000
Epoch 165/1000
```

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Epoch 166/1000
83/83 [============ ] - 10s 119ms/step - loss: 7.1285e-04
Epoch 167/1000
Epoch 168/1000
Epoch 169/1000
Epoch 170/1000
Epoch 171/1000
Epoch 172/1000
Epoch 173/1000
83/83 [============= ] - 10s 119ms/step - loss: 6.6338e-04
Epoch 174/1000
Epoch 175/1000
Epoch 176/1000
Epoch 177/1000
Epoch 178/1000
Epoch 179/1000
Epoch 180/1000
Epoch 181/1000
Epoch 182/1000
Epoch 183/1000
Epoch 184/1000
Epoch 185/1000
Epoch 186/1000
Epoch 187/1000
83/83 [============ ] - 10s 120ms/step - loss: 6.0793e-04
Epoch 188/1000
Epoch 189/1000
```

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Epoch 190/1000
Epoch 191/1000
Epoch 192/1000
83/83 [============== ] - 11s 130ms/step - loss: 6.2439e-04
Epoch 193/1000
Epoch 194/1000
Epoch 195/1000
Epoch 196/1000
Epoch 197/1000
83/83 [============ ] - 10s 118ms/step - loss: 6.7697e-04
Epoch 198/1000
Epoch 199/1000
Epoch 200/1000
Epoch 201/1000
Epoch 202/1000
Epoch 203/1000
Epoch 204/1000
Epoch 205/1000
Epoch 206/1000
83/83 [============== ] - 10s 119ms/step - loss: 6.4589e-04
Epoch 207/1000
Epoch 208/1000
Epoch 209/1000
Epoch 210/1000
Epoch 211/1000
83/83 [============ ] - 10s 120ms/step - loss: 6.8550e-04
Epoch 212/1000
Epoch 213/1000
```

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Epoch 214/1000
Epoch 215/1000
Epoch 216/1000
Epoch 217/1000
Epoch 218/1000
Epoch 219/1000
Epoch 220/1000
Epoch 221/1000
83/83 [============ ] - 10s 126ms/step - loss: 6.4427e-04
Epoch 222/1000
Epoch 223/1000
Epoch 224/1000
Epoch 225/1000
Epoch 226/1000
Epoch 227/1000
Epoch 228/1000
Epoch 229/1000
Epoch 230/1000
Epoch 231/1000
Epoch 232/1000
Epoch 233/1000
Epoch 234/1000
Epoch 235/1000
83/83 [============ ] - 10s 125ms/step - loss: 6.1946e-04
Epoch 236/1000
Epoch 237/1000
```

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Epoch 238/1000
Epoch 239/1000
Epoch 240/1000
83/83 [============== ] - 10s 119ms/step - loss: 6.6018e-04
Epoch 241/1000
Epoch 242/1000
Epoch 243/1000
Epoch 244/1000
Epoch 245/1000
83/83 [============ ] - 10s 120ms/step - loss: 6.3147e-04
Epoch 246/1000
Epoch 247/1000
Epoch 248/1000
Epoch 249/1000
Epoch 250/1000
Epoch 251/1000
Epoch 252/1000
Epoch 253/1000
Epoch 254/1000
Epoch 255/1000
Epoch 256/1000
Epoch 257/1000
Epoch 258/1000
Epoch 259/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.8759e-04
Epoch 260/1000
Epoch 261/1000
```

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Epoch 262/1000
Epoch 263/1000
Epoch 264/1000
Epoch 265/1000
Epoch 266/1000
Epoch 267/1000
Epoch 268/1000
Epoch 269/1000
83/83 [============ ] - 10s 120ms/step - loss: 6.4453e-04
Epoch 270/1000
Epoch 271/1000
Epoch 272/1000
Epoch 273/1000
Epoch 274/1000
Epoch 275/1000
Epoch 276/1000
Epoch 277/1000
Epoch 278/1000
83/83 [============= ] - 10s 118ms/step - loss: 6.1100e-04
Epoch 279/1000
Epoch 280/1000
Epoch 281/1000
Epoch 282/1000
Epoch 283/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.6592e-04
Epoch 284/1000
Epoch 285/1000
```

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Epoch 286/1000
Epoch 287/1000
Epoch 288/1000
Epoch 289/1000
Epoch 290/1000
Epoch 291/1000
Epoch 292/1000
Epoch 293/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.6895e-04
Epoch 294/1000
Epoch 295/1000
Epoch 296/1000
Epoch 297/1000
Epoch 298/1000
Epoch 299/1000
Epoch 300/1000
Epoch 301/1000
Epoch 302/1000
Epoch 303/1000
Epoch 304/1000
Epoch 305/1000
Epoch 306/1000
Epoch 307/1000
83/83 [============ ] - 10s 124ms/step - loss: 5.8089e-04
Epoch 308/1000
Epoch 309/1000
```

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Epoch 310/1000
Epoch 311/1000
Epoch 312/1000
83/83 [============== ] - 12s 147ms/step - loss: 6.1588e-04
Epoch 313/1000
Epoch 314/1000
Epoch 315/1000
Epoch 316/1000
Epoch 317/1000
83/83 [============ ] - 10s 119ms/step - loss: 5.2123e-04
Epoch 318/1000
Epoch 319/1000
Epoch 320/1000
Epoch 321/1000
Epoch 322/1000
Epoch 323/1000
Epoch 324/1000
Epoch 325/1000
Epoch 326/1000
Epoch 327/1000
Epoch 328/1000
Epoch 329/1000
Epoch 330/1000
Epoch 331/1000
83/83 [============ ] - 11s 136ms/step - loss: 5.7373e-04
Epoch 332/1000
Epoch 333/1000
```

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Epoch 334/1000
Epoch 335/1000
Epoch 336/1000
Epoch 337/1000
Epoch 338/1000
Epoch 339/1000
Epoch 340/1000
Epoch 341/1000
83/83 [============ ] - 11s 132ms/step - loss: 5.5762e-04
Epoch 342/1000
Epoch 343/1000
Epoch 344/1000
Epoch 345/1000
Epoch 346/1000
Epoch 347/1000
Epoch 348/1000
Epoch 349/1000
Epoch 350/1000
Epoch 351/1000
Epoch 352/1000
Epoch 353/1000
Epoch 354/1000
Epoch 355/1000
83/83 [============ ] - 14s 165ms/step - loss: 5.6335e-04
Epoch 356/1000
Epoch 357/1000
```

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Epoch 358/1000
Epoch 359/1000
Epoch 360/1000
Epoch 361/1000
Epoch 362/1000
Epoch 363/1000
Epoch 364/1000
Epoch 365/1000
83/83 [============ ] - 11s 135ms/step - loss: 5.8732e-04
Epoch 366/1000
Epoch 367/1000
Epoch 368/1000
Epoch 369/1000
Epoch 370/1000
Epoch 371/1000
Epoch 372/1000
Epoch 373/1000
Epoch 374/1000
Epoch 375/1000
Epoch 376/1000
Epoch 377/1000
Epoch 378/1000
Epoch 379/1000
83/83 [============ ] - 10s 123ms/step - loss: 5.4069e-04
Epoch 380/1000
Epoch 381/1000
```

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Epoch 382/1000
Epoch 383/1000
Epoch 384/1000
Epoch 385/1000
Epoch 386/1000
Epoch 387/1000
Epoch 388/1000
Epoch 389/1000
83/83 [============ ] - 12s 140ms/step - loss: 5.2628e-04
Epoch 390/1000
Epoch 391/1000
Epoch 392/1000
Epoch 393/1000
Epoch 394/1000
Epoch 395/1000
Epoch 396/1000
Epoch 397/1000
Epoch 398/1000
Epoch 399/1000
Epoch 400/1000
Epoch 401/1000
Epoch 402/1000
Epoch 403/1000
83/83 [============ ] - 10s 123ms/step - loss: 5.5210e-04
Epoch 404/1000
Epoch 405/1000
```

```
Epoch 406/1000
83/83 [============ ] - 12s 148ms/step - loss: 4.9851e-04
Epoch 407/1000
Epoch 408/1000
Epoch 409/1000
Epoch 410/1000
Epoch 411/1000
Epoch 412/1000
Epoch 413/1000
83/83 [============ ] - 10s 123ms/step - loss: 5.6088e-04
Epoch 414/1000
Epoch 415/1000
Epoch 416/1000
Epoch 417/1000
Epoch 418/1000
Epoch 419/1000
Epoch 420/1000
Epoch 421/1000
Epoch 422/1000
Epoch 423/1000
Epoch 424/1000
Epoch 425/1000
Epoch 426/1000
Epoch 427/1000
83/83 [============ ] - 12s 144ms/step - loss: 4.8872e-04
Epoch 428/1000
Epoch 429/1000
```

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Epoch 430/1000
Epoch 431/1000
Epoch 432/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.2651e-04
Epoch 433/1000
Epoch 434/1000
Epoch 435/1000
Epoch 436/1000
Epoch 437/1000
83/83 [============ ] - 10s 125ms/step - loss: 5.5888e-04
Epoch 438/1000
Epoch 439/1000
Epoch 440/1000
Epoch 441/1000
Epoch 442/1000
Epoch 443/1000
Epoch 444/1000
Epoch 445/1000
Epoch 446/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.4416e-04
Epoch 447/1000
Epoch 448/1000
Epoch 449/1000
Epoch 450/1000
Epoch 451/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.2119e-04
Epoch 452/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.9112e-04
Epoch 453/1000
```

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Epoch 454/1000
Epoch 455/1000
Epoch 456/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.4710e-04
Epoch 457/1000
Epoch 458/1000
Epoch 459/1000
Epoch 460/1000
Epoch 461/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.1508e-04
Epoch 462/1000
Epoch 463/1000
Epoch 464/1000
Epoch 465/1000
Epoch 466/1000
Epoch 467/1000
Epoch 468/1000
Epoch 469/1000
Epoch 470/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.7206e-04
Epoch 471/1000
Epoch 472/1000
Epoch 473/1000
Epoch 474/1000
Epoch 475/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.5636e-04
Epoch 476/1000
Epoch 477/1000
```

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Epoch 478/1000
Epoch 479/1000
Epoch 480/1000
83/83 [============== ] - 10s 116ms/step - loss: 5.2546e-04
Epoch 481/1000
Epoch 482/1000
Epoch 483/1000
Epoch 484/1000
Epoch 485/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.3450e-04
Epoch 486/1000
Epoch 487/1000
Epoch 488/1000
Epoch 489/1000
Epoch 490/1000
Epoch 491/1000
Epoch 492/1000
Epoch 493/1000
Epoch 494/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.7304e-04
Epoch 495/1000
Epoch 496/1000
Epoch 497/1000
Epoch 498/1000
Epoch 499/1000
83/83 [============ ] - 10s 120ms/step - loss: 5.2584e-04
Epoch 500/1000
Epoch 501/1000
```

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Epoch 502/1000
Epoch 503/1000
Epoch 504/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.9540e-04
Epoch 505/1000
Epoch 506/1000
Epoch 507/1000
Epoch 508/1000
Epoch 509/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.4072e-04
Epoch 510/1000
Epoch 511/1000
Epoch 512/1000
Epoch 513/1000
Epoch 514/1000
Epoch 515/1000
Epoch 516/1000
Epoch 517/1000
Epoch 518/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.1938e-04
Epoch 519/1000
Epoch 520/1000
Epoch 521/1000
Epoch 522/1000
Epoch 523/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.4809e-04
Epoch 524/1000
Epoch 525/1000
```

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Epoch 526/1000
Epoch 527/1000
Epoch 528/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.2178e-04
Epoch 529/1000
Epoch 530/1000
Epoch 531/1000
Epoch 532/1000
Epoch 533/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.1960e-04
Epoch 534/1000
Epoch 535/1000
Epoch 536/1000
Epoch 537/1000
Epoch 538/1000
Epoch 539/1000
Epoch 540/1000
Epoch 541/1000
Epoch 542/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.0348e-04
Epoch 543/1000
Epoch 544/1000
Epoch 545/1000
Epoch 546/1000
Epoch 547/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.1378e-04
Epoch 548/1000
Epoch 549/1000
```

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Epoch 550/1000
Epoch 551/1000
Epoch 552/1000
Epoch 553/1000
Epoch 554/1000
Epoch 555/1000
Epoch 556/1000
Epoch 557/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0231e-04
Epoch 558/1000
Epoch 559/1000
Epoch 560/1000
Epoch 561/1000
Epoch 562/1000
Epoch 563/1000
Epoch 564/1000
Epoch 565/1000
Epoch 566/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.6780e-04
Epoch 567/1000
Epoch 568/1000
Epoch 569/1000
Epoch 570/1000
Epoch 571/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.2426e-04
Epoch 572/1000
Epoch 573/1000
```

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Epoch 574/1000
Epoch 575/1000
Epoch 576/1000
Epoch 577/1000
Epoch 578/1000
Epoch 579/1000
Epoch 580/1000
Epoch 581/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.2695e-04
Epoch 582/1000
Epoch 583/1000
Epoch 584/1000
Epoch 585/1000
Epoch 586/1000
Epoch 587/1000
Epoch 588/1000
Epoch 589/1000
Epoch 590/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.1428e-04
Epoch 591/1000
Epoch 592/1000
Epoch 593/1000
Epoch 594/1000
Epoch 595/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0854e-04
Epoch 596/1000
Epoch 597/1000
```

```
Epoch 598/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.0518e-04
Epoch 599/1000
Epoch 600/1000
Epoch 601/1000
Epoch 602/1000
Epoch 603/1000
Epoch 604/1000
Epoch 605/1000
83/83 [============ ] - 10s 118ms/step - loss: 5.2137e-04
Epoch 606/1000
Epoch 607/1000
Epoch 608/1000
Epoch 609/1000
Epoch 610/1000
Epoch 611/1000
Epoch 612/1000
Epoch 613/1000
Epoch 614/1000
83/83 [============== ] - 11s 131ms/step - loss: 5.3615e-04
Epoch 615/1000
Epoch 616/1000
Epoch 617/1000
Epoch 618/1000
Epoch 619/1000
83/83 [============ ] - 10s 120ms/step - loss: 5.1998e-04
Epoch 620/1000
Epoch 621/1000
```

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Epoch 622/1000
Epoch 623/1000
Epoch 624/1000
Epoch 625/1000
Epoch 626/1000
Epoch 627/1000
Epoch 628/1000
Epoch 629/1000
83/83 [============ ] - 10s 125ms/step - loss: 5.3019e-04
Epoch 630/1000
Epoch 631/1000
Epoch 632/1000
Epoch 633/1000
Epoch 634/1000
Epoch 635/1000
Epoch 636/1000
Epoch 637/1000
Epoch 638/1000
83/83 [============== ] - 10s 125ms/step - loss: 5.3134e-04
Epoch 639/1000
Epoch 640/1000
Epoch 641/1000
Epoch 642/1000
Epoch 643/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.9557e-04
Epoch 644/1000
Epoch 645/1000
```

```
Epoch 646/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8572e-04
Epoch 647/1000
Epoch 648/1000
83/83 [============== ] - 10s 116ms/step - loss: 4.8401e-04
Epoch 649/1000
Epoch 650/1000
Epoch 651/1000
Epoch 652/1000
Epoch 653/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.0051e-04
Epoch 654/1000
Epoch 655/1000
Epoch 656/1000
Epoch 657/1000
Epoch 658/1000
Epoch 659/1000
Epoch 660/1000
Epoch 661/1000
Epoch 662/1000
83/83 [============== ] - 10s 118ms/step - loss: 5.2878e-04
Epoch 663/1000
Epoch 664/1000
Epoch 665/1000
Epoch 666/1000
Epoch 667/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8265e-04
Epoch 668/1000
Epoch 669/1000
```

```
Epoch 670/1000
Epoch 671/1000
Epoch 672/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.1478e-04
Epoch 673/1000
Epoch 674/1000
Epoch 675/1000
Epoch 676/1000
Epoch 677/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.9702e-04
Epoch 678/1000
Epoch 679/1000
Epoch 680/1000
Epoch 681/1000
Epoch 682/1000
Epoch 683/1000
Epoch 684/1000
Epoch 685/1000
Epoch 686/1000
83/83 [============= ] - 10s 116ms/step - loss: 4.5679e-04
Epoch 687/1000
Epoch 688/1000
Epoch 689/1000
Epoch 690/1000
Epoch 691/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8345e-04
Epoch 692/1000
Epoch 693/1000
```

```
Epoch 694/1000
83/83 [============ ] - 10s 117ms/step - loss: 5.2664e-04
Epoch 695/1000
Epoch 696/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.0196e-04
Epoch 697/1000
Epoch 698/1000
Epoch 699/1000
Epoch 700/1000
Epoch 701/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.6609e-04
Epoch 702/1000
Epoch 703/1000
Epoch 704/1000
Epoch 705/1000
Epoch 706/1000
Epoch 707/1000
Epoch 708/1000
Epoch 709/1000
Epoch 710/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.1417e-04
Epoch 711/1000
Epoch 712/1000
Epoch 713/1000
Epoch 714/1000
Epoch 715/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8268e-04
Epoch 716/1000
Epoch 717/1000
```

```
Epoch 718/1000
Epoch 719/1000
Epoch 720/1000
83/83 [============== ] - 10s 117ms/step - loss: 5.0007e-04
Epoch 721/1000
Epoch 722/1000
Epoch 723/1000
Epoch 724/1000
Epoch 725/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.7315e-04
Epoch 726/1000
Epoch 727/1000
Epoch 728/1000
Epoch 729/1000
Epoch 730/1000
Epoch 731/1000
Epoch 732/1000
Epoch 733/1000
Epoch 734/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.8700e-04
Epoch 735/1000
Epoch 736/1000
Epoch 737/1000
Epoch 738/1000
Epoch 739/1000
83/83 [============ ] - 11s 134ms/step - loss: 5.3182e-04
Epoch 740/1000
Epoch 741/1000
```

```
Epoch 742/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.7992e-04
Epoch 743/1000
Epoch 744/1000
Epoch 745/1000
Epoch 746/1000
Epoch 747/1000
Epoch 748/1000
Epoch 749/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.9415e-04
Epoch 750/1000
Epoch 751/1000
Epoch 752/1000
Epoch 753/1000
Epoch 754/1000
Epoch 755/1000
Epoch 756/1000
Epoch 757/1000
Epoch 758/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.5592e-04
Epoch 759/1000
Epoch 760/1000
Epoch 761/1000
Epoch 762/1000
Epoch 763/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8082e-04
Epoch 764/1000
Epoch 765/1000
```

```
Epoch 766/1000
Epoch 767/1000
Epoch 768/1000
Epoch 769/1000
Epoch 770/1000
Epoch 771/1000
Epoch 772/1000
Epoch 773/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.8768e-04
Epoch 774/1000
Epoch 775/1000
Epoch 776/1000
Epoch 777/1000
Epoch 778/1000
Epoch 779/1000
Epoch 780/1000
Epoch 781/1000
Epoch 782/1000
83/83 [============= ] - 10s 117ms/step - loss: 4.8000e-04
Epoch 783/1000
Epoch 784/1000
Epoch 785/1000
Epoch 786/1000
Epoch 787/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.5171e-04
Epoch 788/1000
Epoch 789/1000
```

```
Epoch 790/1000
Epoch 791/1000
Epoch 792/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.8716e-04
Epoch 793/1000
Epoch 794/1000
Epoch 795/1000
Epoch 796/1000
Epoch 797/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.5988e-04
Epoch 798/1000
Epoch 799/1000
Epoch 800/1000
Epoch 801/1000
Epoch 802/1000
Epoch 803/1000
Epoch 804/1000
Epoch 805/1000
Epoch 806/1000
Epoch 807/1000
Epoch 808/1000
Epoch 809/1000
Epoch 810/1000
Epoch 811/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.5517e-04
Epoch 812/1000
Epoch 813/1000
```

```
Epoch 814/1000
Epoch 815/1000
Epoch 816/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.4188e-04
Epoch 817/1000
Epoch 818/1000
Epoch 819/1000
Epoch 820/1000
Epoch 821/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.7413e-04
Epoch 822/1000
Epoch 823/1000
Epoch 824/1000
Epoch 825/1000
Epoch 826/1000
Epoch 827/1000
Epoch 828/1000
Epoch 829/1000
Epoch 830/1000
Epoch 831/1000
Epoch 832/1000
Epoch 833/1000
Epoch 834/1000
Epoch 835/1000
83/83 [============ ] - 10s 126ms/step - loss: 4.5315e-04
Epoch 836/1000
Epoch 837/1000
```

```
Epoch 838/1000
Epoch 839/1000
Epoch 840/1000
Epoch 841/1000
Epoch 842/1000
Epoch 843/1000
Epoch 844/1000
Epoch 845/1000
83/83 [============ ] - 10s 123ms/step - loss: 4.6172e-04
Epoch 846/1000
Epoch 847/1000
Epoch 848/1000
Epoch 849/1000
Epoch 850/1000
Epoch 851/1000
Epoch 852/1000
Epoch 853/1000
Epoch 854/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.9466e-04
Epoch 855/1000
Epoch 856/1000
Epoch 857/1000
Epoch 858/1000
Epoch 859/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.6931e-04
Epoch 860/1000
Epoch 861/1000
```

```
Epoch 862/1000
Epoch 863/1000
Epoch 864/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.5627e-04
Epoch 865/1000
Epoch 866/1000
Epoch 867/1000
Epoch 868/1000
Epoch 869/1000
83/83 [=========== ] - 10s 117ms/step - loss: 4.6090e-04
Epoch 870/1000
Epoch 871/1000
Epoch 872/1000
Epoch 873/1000
Epoch 874/1000
Epoch 875/1000
Epoch 876/1000
Epoch 877/1000
Epoch 878/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.4354e-04
Epoch 879/1000
Epoch 880/1000
Epoch 881/1000
Epoch 882/1000
Epoch 883/1000
83/83 [============ ] - 10s 116ms/step - loss: 4.2977e-04
Epoch 884/1000
Epoch 885/1000
```

```
Epoch 886/1000
Epoch 887/1000
Epoch 888/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.5549e-04
Epoch 889/1000
Epoch 890/1000
Epoch 891/1000
Epoch 892/1000
Epoch 893/1000
83/83 [=========== ] - 10s 117ms/step - loss: 4.0740e-04
Epoch 894/1000
Epoch 895/1000
Epoch 896/1000
Epoch 897/1000
Epoch 898/1000
Epoch 899/1000
Epoch 900/1000
Epoch 901/1000
Epoch 902/1000
Epoch 903/1000
Epoch 904/1000
Epoch 905/1000
Epoch 906/1000
Epoch 907/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.2727e-04
Epoch 908/1000
Epoch 909/1000
```

```
Epoch 910/1000
Epoch 911/1000
Epoch 912/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.4110e-04
Epoch 913/1000
Epoch 914/1000
Epoch 915/1000
Epoch 916/1000
Epoch 917/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.7186e-04
Epoch 918/1000
Epoch 919/1000
Epoch 920/1000
Epoch 921/1000
Epoch 922/1000
Epoch 923/1000
Epoch 924/1000
Epoch 925/1000
Epoch 926/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.5034e-04
Epoch 927/1000
Epoch 928/1000
Epoch 929/1000
Epoch 930/1000
Epoch 931/1000
83/83 [============ ] - 10s 119ms/step - loss: 4.4962e-04
Epoch 932/1000
Epoch 933/1000
```

```
Epoch 934/1000
Epoch 935/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.6672e-04
Epoch 936/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.4030e-04
Epoch 937/1000
Epoch 938/1000
Epoch 939/1000
Epoch 940/1000
Epoch 941/1000
83/83 [=========== ] - 10s 118ms/step - loss: 4.3226e-04
Epoch 942/1000
Epoch 943/1000
Epoch 944/1000
Epoch 945/1000
Epoch 946/1000
Epoch 947/1000
Epoch 948/1000
Epoch 949/1000
Epoch 950/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.4070e-04
Epoch 951/1000
Epoch 952/1000
Epoch 953/1000
Epoch 954/1000
Epoch 955/1000
83/83 [============ ] - 10s 117ms/step - loss: 4.3441e-04
Epoch 956/1000
Epoch 957/1000
```

```
Epoch 958/1000
Epoch 959/1000
83/83 [============= ] - 10s 118ms/step - loss: 4.7068e-04
Epoch 960/1000
83/83 [============== ] - 10s 117ms/step - loss: 4.6130e-04
Epoch 961/1000
Epoch 962/1000
Epoch 963/1000
Epoch 964/1000
Epoch 965/1000
83/83 [============ ] - 10s 118ms/step - loss: 4.2591e-04
Epoch 966/1000
Epoch 967/1000
Epoch 968/1000
Epoch 969/1000
Epoch 970/1000
Epoch 971/1000
Epoch 972/1000
Epoch 973/1000
Epoch 974/1000
83/83 [============== ] - 10s 118ms/step - loss: 4.5958e-04
Epoch 975/1000
Epoch 976/1000
Epoch 977/1000
Epoch 978/1000
Epoch 979/1000
83/83 [============ ] - 10s 117ms/step - loss: 3.9937e-04
Epoch 980/1000
Epoch 981/1000
```

```
Epoch 982/1000
 Epoch 983/1000
 Epoch 984/1000
 Epoch 985/1000
 Epoch 986/1000
 Epoch 987/1000
 Epoch 988/1000
 Epoch 989/1000
 83/83 [============ ] - 10s 123ms/step - loss: 4.1693e-04
 Epoch 990/1000
 Epoch 991/1000
 Epoch 992/1000
 Epoch 993/1000
 Epoch 994/1000
 Epoch 995/1000
 Epoch 996/1000
 Epoch 997/1000
 Epoch 998/1000
 Epoch 999/1000
 Epoch 1000/1000
 [116]: <tensorflow.python.keras.callbacks.History at 0x7fde9518ac10>
[117]: testdataframe= test
  testdataframe['Date'] = testdataframe.index
  testdata = pd.DataFrame(columns = ['Date', 'Close'])
  testdata['Date'] = testdataframe['Date']
  testdata['Close'] = testdataframe['Close']
```

```
real_stock_price = testdata.iloc[:, 1:2].values
dataset_total = pd.concat((data2['Close'], testdata['Close']), axis = 0)
inputs = dataset_total[len(dataset_total) - len(testdata) - 60:].values
inputs = inputs.reshape(-1,1)
inputs = sc.transform(inputs)
X_test = []
for i in range(60, inputs.shape[0]):
    X_test.append(inputs[i-60:i, 0])
X_test = np.array(X_test)
X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], 1))
```

```
[118]: predicted_with_gru = regressorGRU.predict(X_test)
predicted_with_gru = sc.inverse_transform(predicted_with_gru)
```

```
[119]: plt.figure(figsize=(20,10))
   plt.plot(real_stock_price, color = 'green', label = 'TSX Stock Price')
   plt.plot(predicted_with_gru, color = 'red', label = 'Predicted TSX Stock Price')
   plt.title('TSX Stock Price Prediction')
   plt.xlabel('Date')
   plt.ylabel('TSX Stock Price')
   plt.legend()
   plt.show()
```



```
[120]: rmse_predict1= np.reshape(predicted_with_gru,533)
[121]: #forecast metrics
    def smape(a, f):
        return 1/len(a) * np.sum(2 * np.abs(f-a) / (np.abs(a) + np.abs(f))*100)
```

```
smape(test["Close"].values,rmse_predict1)
[121]: 0.5143041587306517
[122]: #forecast metrics
       from sklearn.metrics import mean_absolute_error
       from sklearn.metrics import mean_squared_error
       from sklearn.metrics import r2_score
       mae = mean_absolute_error(test["Close"].values, rmse_predict1)
       mse = mean_squared_error(test["Close"].values, rmse_predict1)
       rmse = np.sqrt(mse)
       print("Results of sklearn.metrics:")
       print("MAE:",mae)
       print("MSE:", mse)
       print("RMSE:", rmse)
      Results of sklearn.metrics:
      MAE: 81.20319763660898
      MSE: 10585.447468399412
      RMSE: 102.88560379566916
[123]: elapsed_time = time.process_time() - t
       print(elapsed_time)
      30051.340034676992
  []:
  []:
[124]: | #upload data to understand the relationship between canada's new covid cases
       \rightarrow and TSX price
       #split data into 2 sections, test -> recent data from sept 8th onwards, train_
        \rightarrow-> feb 1 to march 23rd
       df=pd.read_csv("Canada Covid New Cases & Stock Price.csv", sep=",")
       from datetime import datetime
       con=df['Date']
       df['Date']=pd.to_datetime(df['Date'])
       df.set_index('Date', inplace=True)
       test = df[150:]
       train = df[:35]
[125]: import seaborn as sb
```

```
[126]: # when covid first started around feb, there was a relatively strong negative
        →relationship between covid new cases vs stock price
       pearsoncorr = train.corr(method='pearson')
       pearsoncorr
[126]:
                         Sum of new_cases TSX Price
                                 1.000000 -0.677857
       Sum of new cases
      TSX Price
                                -0.677857
                                            1.000000
[127]: # as covid impact continues, there still is a equtive relationship between
        →covid new cases vs stock price, however the relationship is not as strong as
       → the beginning of covid
       pearsoncorr = test.corr(method='pearson')
       pearsoncorr
[127]:
                         Sum of new cases TSX Price
                                 1.000000 -0.340568
       Sum of new cases
       TSX Price
                                -0.340568
                                           1.000000
[128]: | #upload data to understand the relationship between US's new covid cases and
       \rightarrow NASDAQ price
       #split data into 2 sections, test -> recent data from sept 8th onwards, train_
       \rightarrow-> feb 1 to march 23rd
       df=pd.read_csv("USA Covid New Cases & Stock Price.csv", sep=",")
       from datetime import datetime
       con=df['Date']
       df['Date']=pd.to_datetime(df['Date'])
       df.set_index('Date', inplace=True)
       test = df[150:]
       train = df[:35]
[129]: | # when covid first started around feb, there was a relatively strong negative,
       →relationship between covid new cases vs stock price
       pearsoncorr = train.corr(method='pearson')
       pearsoncorr
[129]:
                         Sum of new_cases Stock Price
       Sum of new cases
                                 1.000000
                                             -0.653411
       Stock Price
                                -0.653411
                                              1.000000
[130]: | # as covid impact continues, there still is a equtive relationship between
       →covid new cases vs stock price, however the relationship is not as strong as
       → the beginning of covid. The relationship is weaker in the US than Canada.
       pearsoncorr = test.corr(method='pearson')
       pearsoncorr
```

[130]:	Sum of new_cases Stock Price	Sum of new_cases 1.000000 -0.168529	Stock Price -0.168529 1.000000	
[]:				
[]:				
[]:				