

9. DEVELOP NEURAL NETWORK-BASED TIME SERIES FORECASTING MODEL

AIM:

To develop Neural Network Based Time Series Forecasting Model.

PROCEDURE:

Step 1: Create synthetic petrol price data

```
np.random.seed(42)
```

```
dates = pd.date_range(start='2015-01-01', periods=120, freq='M') #  
10 years of monthly data
```

```
trend = np.linspace(50, 100, 120)
```

```
seasonal = 5 * np.sin(2 * np.pi * dates.month / 12)
```

```
noise = np.random.normal(0, 2, 120)
```

```
prices = trend + seasonal + noise
```

```
df = pd.DataFrame({'Date': dates, 'Petrol_Price': prices})
```

```
df.set_index('Date', inplace=True)
```

Step 2: Normalize the data

```
scaler = MinMaxScaler()
```

```
scaled_prices = scaler.fit_transform(df[['Petrol_Price']])
```

Step 3: Prepare sequences for LSTM (sliding window)

```
def create_dataset(data, window_size):
```

```
    X, y = [], []
```

```
    for i in range(len(data) - window_size):
```

```
        X.append(data[i:i + window_size])
```

```
        y.append(data[i + window_size])
```

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

```
window_size = 12
```

```
X, y = create_dataset(scaled_prices, window_size)
```

Reshape input for LSTM [samples, time steps, features]

```
X = X.reshape((X.shape[0], X.shape[1], 1))
```

Step 4: Define the LSTM model

```
model = Sequential([
```

```
    LSTM(50, activation='relu', input_shape=(window_size, 1)),
```

```
    Dense(1)
```

```
])
```

```
model.compile(optimizer='adam', loss='mse')
```

Step 5: Train the model

```
model.fit(X, y, epochs=100, verbose=0)
```

```
# Step 6: Forecast future prices
```

```
forecast_input = scaled_prices[-window_size:] # last window
```

```
predictions = []
```

```
n_forecast = 12
```

```
for _ in range(n_forecast):
```

```
    input_resaped = forecast_input.reshape((1, window_size, 1))
```

```
    pred = model.predict(input_resaped, verbose=0)
```

```
    predictions.append(pred[0, 0])
```

```
    forecast_input = np.append(forecast_input[1:], pred[0, 0])
```

```
# Inverse transform to get actual price
```

```
forecasted_prices =
```

```
scaler.inverse_transform(np.array(predictions).reshape(-1, 1))
```

```
# Step 7: Plot
```

```
forecast_dates = pd.date_range(df.index[-1] +
```

```
pd.DateOffset(months=1), periods=n_forecast, freq='M')
```

```
plt.figure(figsize=(10, 4))
```

```
plt.plot(df.index, df['Petrol_Price'], label='Historical Prices')
```

```
plt.plot(forecast_dates, forecasted_prices, color='red',  
label='Forecasted Prices')
```

```
plt.title("LSTM Forecast of Petrol Prices")
```

```
plt.xlabel("Date")
```

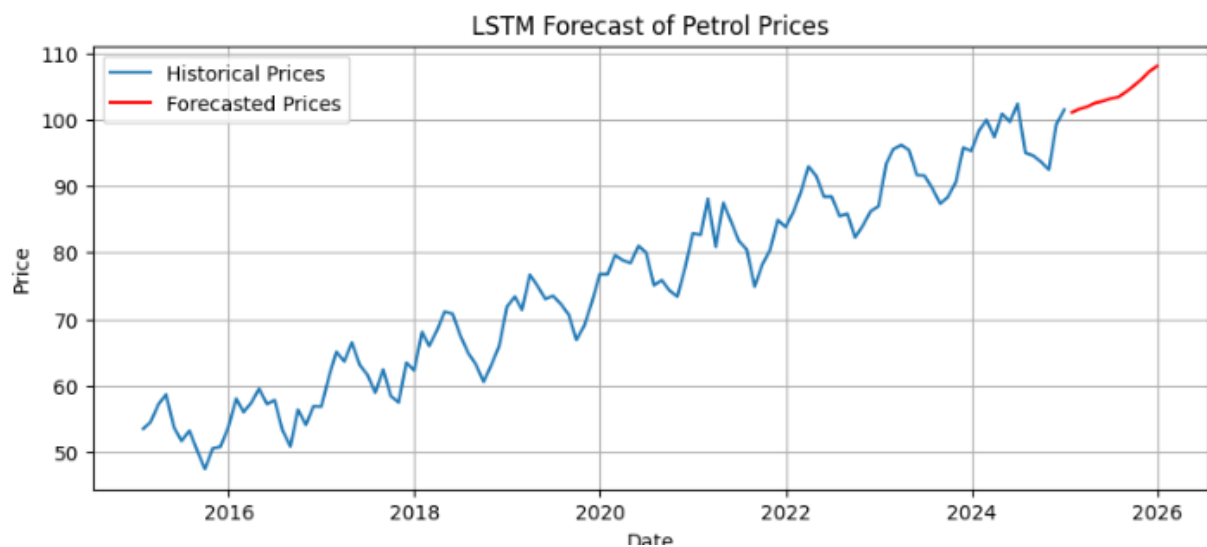
```
plt.ylabel("Price")
```

```
plt.legend()
```

```
plt.grid(True)
```

```
plt.show()
```

OUTPUT:



RESULT:

The program to execute Neural Network-Based Time Series Forecasting Model has been executed successfully.