EX:No.9	
DATE:12/04/25	Develop neural network-based time series forecasting model.

#### AIM:

To Develop neural network-based time series forecasting model.

## **ALGORITHM:**

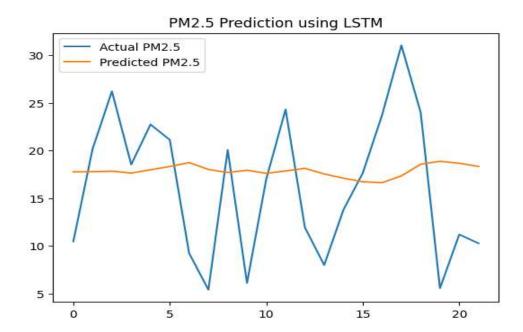
- 1. **Data Cleaning** Loaded the dataset, parsed dates, fixed encoding issues, and selected only the PM2.5 column.
- 2. **Normalization** Scaled PM2.5 values between 0 and 1 using MinMaxScaler to improve neural network performance.
- 3. **Sequence Creation** Created supervised learning format by using the previous 10 timesteps to predict the next one.
- 4. **Train-Test Split** Split the dataset into 80% training and 20% testing sets.
- 5. **Model Building** Built an LSTM model with one LSTM layer (50 units) and one Dense output layer.
- 6. **Model Training** Trained the model using training data for 20 epochs with a batch size of 32.
- 7. **Prediction & Inverse Scaling** Predicted future values and converted them back to original scale.
- 8. **Visualization** Plotted actual vs predicted PM2.5 values to evaluate model performance.

#### Code:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense
# Load dataset
df = pd.read_csv('/content/us_air_pollution_2012_2021.csv', parse_dates=['Date'])
df.set_index('Date', inplace=True)
# Clean and select only PM2.5
df.columns = [col.replace("Â", "") for col in df.columns]
data = df[['PM2.5 (\mu g/m^3)']].dropna()
# Normalize the data
scaler = MinMaxScaler()
scaled = scaler.fit transform(data)
# Convert to supervised format
X, y = [], []
```

```
for i in range(10, len(scaled)):
  X.append(scaled[i-10:i])
  y.append(scaled[i])
X, y = np.array(X), np.array(y)
# Split into train and test
split = int(len(X) * 0.8)
X_train, X_test = X[:split], X[split:]
y_train, y_test = y[:split], y[split:]
# Build model
model = Sequential()
model.add(LSTM(50, activation='relu', input_shape=(10, 1)))
model.add(Dense(1))
model.compile(optimizer='adam', loss='mse')
# Train model
model.fit(X train, y train, epochs=20, batch size=32, validation data=(X test, y test))
# Predict and inverse scale
pred = model.predict(X_test)
pred = scaler.inverse_transform(pred)
actual = scaler.inverse_transform(y_test)
# Plot
plt.plot(actual, label='Actual PM2.5')
plt.plot(pred, label='Predicted PM2.5')
plt.legend()
plt.title('PM2.5 Prediction using LSTM')
plt.show()
```

### **OUTPUT:**



# **RESULT:**

Thus, the program using the time series data implementation has been done successfully.

