

5. Develop a linear regression model for forecasting time series data

AIM:

To develop a **Linear Regression model** to **forecast future values** based on historical **time series** data from an expenses dataset.

PROCEDURE:

- 1) Import libraries (like pandas, matplotlib, sklearn, etc.)
- 2) Load the dataset (expenses.csv).
- 3) Explore the dataset (check columns, types, missing values).
- 4) Convert date column (if present) to datetime type.
- 5) Set date as index (common practice in time series).
- 6) Feature Engineering: Create a numerical time variable (for linear regression).
- 7) Split the data into training and testing sets.
- 8) Train the Linear Regression model.
- 9) Predict on test data.
- 10) Visualize the results (Actual vs Predicted).
- 11) Forecast future points.

CODE:

```
# Step 1: Import Libraries
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.linear_model import LinearRegression
```

```
from sklearn.metrics import mean_squared_error
```

```
# Step 2: Load Dataset
```

```
data = pd.read_csv('/mnt/data/expenses.csv')
```

```
# Step 3: Explore Dataset
```

```
print(data.head())
```

```
print(data.info())
```

```
# Step 4: Convert 'Date' column to datetime
```

```
data['Date'] = pd.to_datetime(data['Date'])
```

```
# Step 5: Set Date as Index
```

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

```
# Step 6: Feature Engineering
```

```
# Create a numerical feature for time
```

```
data['Time'] = range(len(data))
```

```
# Step 7: Prepare features and target
```

```
X = data[['Time']]
```

```
y = data['Expenses']
```

```
# Step 8: Train-Test Split (80% train, 20% test)
```

```
split = int(0.8 * len(data))
```

```
X_train, X_test = X[:split], X[split:]
```

```
y_train, y_test = y[:split], y[split:]
```

```
# Step 9: Train the Linear Regression Model
```

```
model = LinearRegression()
```

```
model.fit(X_train, y_train)
```

```
# Step 10: Predict
```

```
y_pred = model.predict(X_test)
```

```
# Step 11: Evaluate the Model
```

```
mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error: {mse}")
```

```
# Step 12: Plot Actual vs Predicted
```

```
plt.figure(figsize=(10,6))
plt.plot(data.index[split:], y_test, label='Actual', marker='o')
plt.plot(data.index[split:], y_pred, label='Predicted', marker='x')
plt.title('Actual vs Predicted Expenses')
plt.xlabel('Date')
plt.ylabel('Expenses')
plt.legend()
plt.grid(True)
plt.show()
```

```
# Step 13: Forecast Future
```

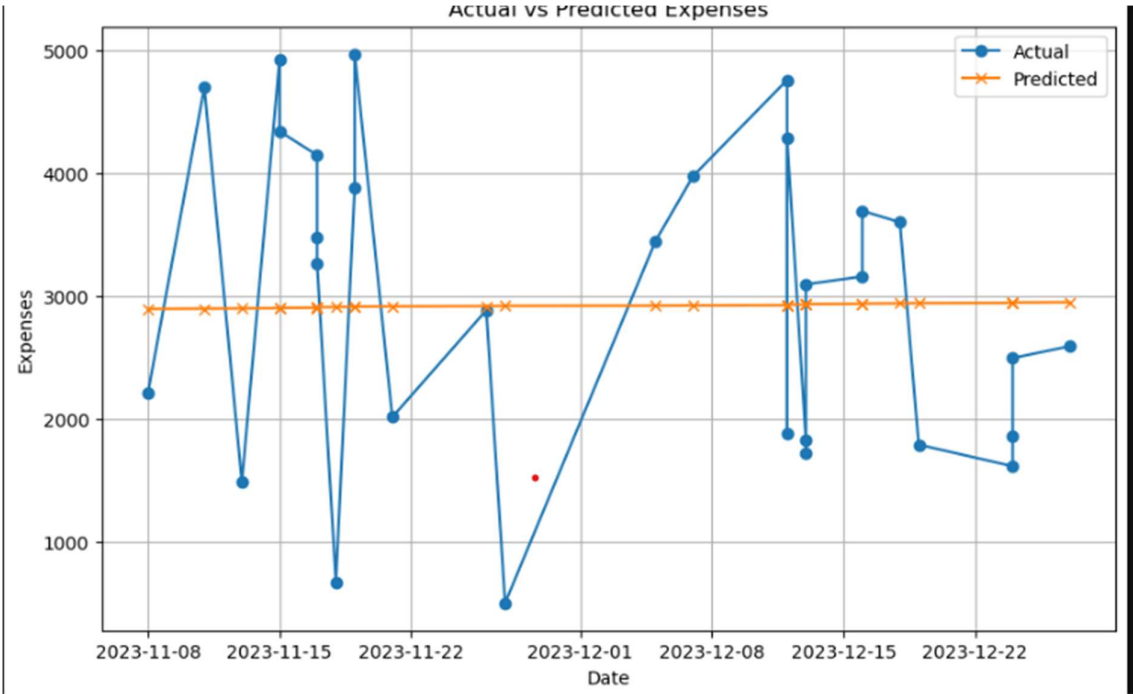
```
# Forecast next 10 time points
```

```
future_time = pd.DataFrame({'Time': range(len(data), len(data)+10)})
future_pred = model.predict(future_time)
```

```
# Display forecasted values
```

```
print("Forecasted Future Expenses:")
print(future_pred)
```

OUTPUT:



RESULT:

The program to develop a linear regression model is implemented successfully.