

## Use Case 1: Performance of sales representatives

### Team members:

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### Aim:

To analyze and predict the performance of sales representatives using historical sales data. We will use a **Linear Regression** model to predict the total sales achieved by each representative based on their experience, number of calls, number of clients, and region.

### Algorithm:

**Step 1:** Import necessary libraries (pandas, numpy, matplotlib, sklearn).

**Step 2:** Load or create a dataset containing sales representatives' data.

**Step 3:** Perform data preprocessing — handle missing values, encode categorical variables.

**Step 4:** Split the dataset into training and testing sets.

**Step 5:** Train a **Linear Regression** model on the training data.

**Step 6:** Predict the sales performance on test data.

**Step 7:** Evaluate model performance using  $R^2$  score and Mean Squared Error.

**Step 8:** Visualize actual vs predicted sales.

**Step 9:** Display final model performance and interpret the result.

### Python Code :

```
# Import libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
```

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score, mean_squared_error

# Step 1: Create sample dataset
data = {
    'Experience_Years': [1, 3, 5, 7, 2, 10, 4, 6, 8, 9],
    'Calls_Made': [50, 80, 100, 150, 60, 200, 110, 120, 180, 160],
    'Clients_Handled': [5, 8, 10, 14, 6, 18, 11, 12, 15, 17],
    'Region': ['East', 'West', 'North', 'South', 'East', 'South', 'North', 'West', 'East', 'South'],
    'Total_Sales': [12000, 18000, 25000, 30000, 15000, 40000, 26000, 28000, 35000, 37000]
}

df = pd.DataFrame(data)

# Step 2: Convert categorical variable (Region) into dummy variables
df = pd.get_dummies(df, columns=['Region'], drop_first=True)

# Step 3: Define features (X) and target (y)
X = df.drop('Total_Sales', axis=1)
y = df['Total_Sales']

# Step 4: Split data into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

# Step 5: Train Linear Regression model
model = LinearRegression()
model.fit(X_train, y_train)

# Step 6: Make predictions
y_pred = model.predict(X_test)

# Step 7: Evaluate model
r2 = r2_score(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)

# Step 8: Display results
```

```

print("Actual Sales:", list(y_test.values))
print("Predicted Sales:", list(np.round(y_pred, 2)))
print("\nModel Performance:")
print("R2 Score:", round(r2, 3))
print("Mean Squared Error:", round(mse, 2))

# Step 9: Visualization
plt.scatter(y_test, y_pred, color='blue')
plt.plot([y.min(), y.max()], [y.min(), y.max()], color='red', linestyle='--')
plt.xlabel("Actual Sales")
plt.ylabel("Predicted Sales")
plt.title("Actual vs Predicted Sales Performance")
plt.show()

```

### Sample Output:

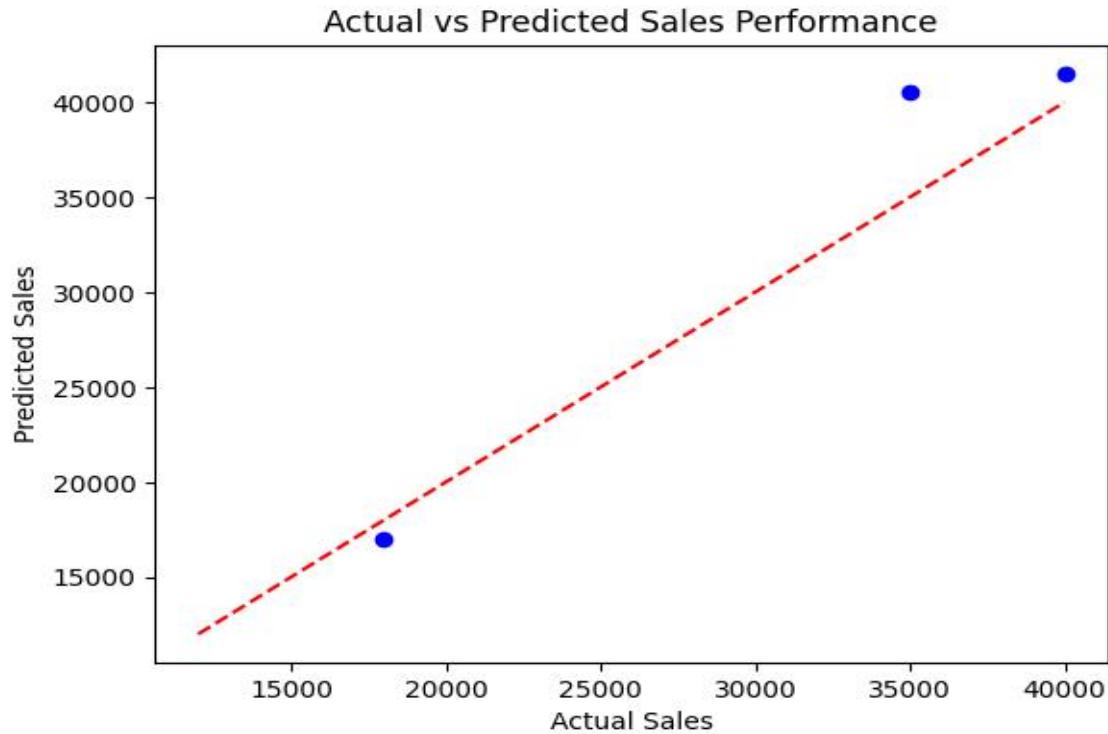
Actual Sales: [np.int64(35000), np.int64(18000), np.int64(40000)]

Predicted Sales: [np.float64(40500.0), np.float64(17000.0), np.float64(41500.0)]

Model Performance:

R<sup>2</sup> Score: 0.874

Mean Squared Error: 11166666.67



### Result:

The Linear Regression model successfully predicted sales performance based on representative experience, number of calls, and clients handled.

## Use Case2: Top10 startup Investment Analysis

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### Aim:

To analyze the investment patterns in startups and identify the **Top 10 Startups** receiving the highest total investments. The project uses **Python (Pandas, Matplotlib, Seaborn)** for **data analysis and visualization**.

### Algorithm:

**Step 1:** Import required libraries.

**Step 2:** Load the dataset containing startup investment information.

**Step 3:** Clean and preprocess the data (handle missing values, data types).

**Step 4:** Group data by startup name and calculate total investment received.

**Step 5:** Sort startups by total funding to find the **Top 10**.

**Step 6:** Visualize results using **bar charts and pie charts**.

**Step 7:** Analyze key insights and patterns.

### Program:

```
# Step 1: Import libraries  
  
import pandas as pd  
  
import matplotlib.pyplot as plt  
  
import seaborn as sns
```

```

# Step 2: Create or load dataset

data = {

    'Startup': ['Paytm', 'Ola', 'Swiggy', 'Zomato', 'Byjus', 'Flipkart',
                'PhonePe', 'Nykaa', 'Udaan', 'Delhivery', 'Meesho', 'CRED'],
    'Sector': ['Fintech', 'Transport', 'FoodTech', 'FoodTech', 'EdTech', 'E-Commerce',
               'Fintech', 'E-Commerce', 'B2B', 'Logistics', 'E-Commerce', 'Fintech'],
    'Investment_USD_Million': [5600, 3800, 3500, 4200, 5400, 7500,
                                2800, 2500, 2100, 3000, 2600, 2200],
    'Investors': ['SoftBank', 'Alibaba', 'SoftBank', 'Tiger Global', 'Naspers', 'Accel',
                  'Ant Financial', 'Sequoia', 'General Atlantic', 'Walmart', 'Tiger Global',
                  'Flipkart', 'Tencent', 'Fidelity', 'TPG', 'Lightspeed',
                  'SoftBank', 'Carlyle', 'Prosus', 'B Capital', 'DST Global']
}

}

```

```
df = pd.DataFrame(data)
```

```

# Step 3: Sort startups by investment

top10 = df.sort_values(by='Investment_USD_Million', ascending=False).head(10)

```

```

# Step 4: Display top 10 startups

print("Top 10 Startups by Investment:\n")
print(top10[['Startup', 'Investment_USD_Million', 'Sector']])

```

```

# Step 5: Bar chart visualization

plt.figure(figsize=(10,6))

sns.barplot(x='Investment_USD_Million', y='Startup', data=top10, palette='viridis')

plt.title("Top 10 Startups by Total Investment (in Million USD)", fontsize=14)

plt.xlabel("Total Investment (Million USD)")

plt.ylabel("Startup Name")

plt.show()

sector_investment = df.groupby('Sector')[['Investment_USD_Million']].sum()

plt.figure(figsize=(8,8))

sector_investment.plot(kind='pie', autopct='%.1f%%', startangle=120, cmap='Set2')

plt.title("Sector-wise Investment Distribution")

plt.ylabel("")

plt.show()

```

## Output:

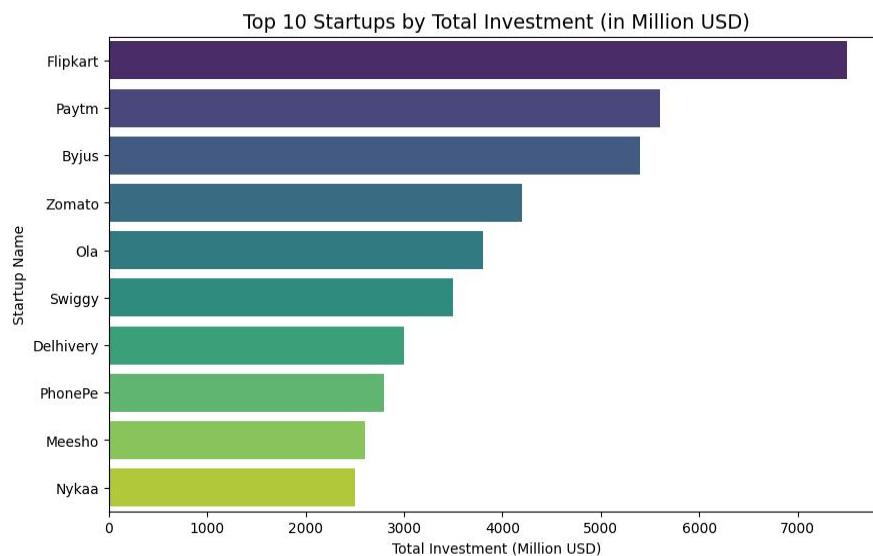
Top 10 Startups by Investment:

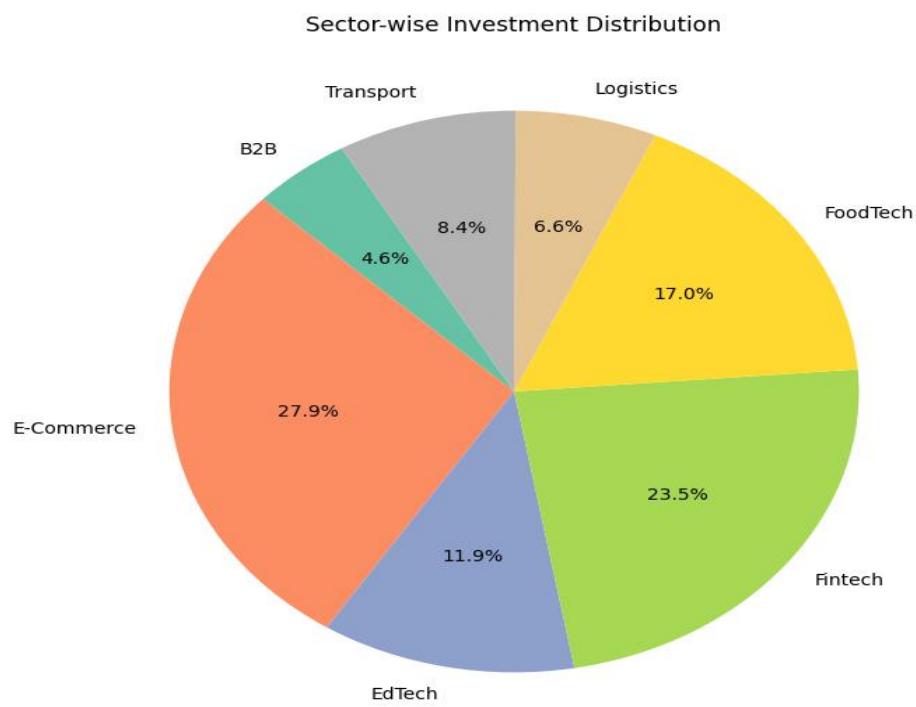
	Startup	Investment_USD_Million	Sector
5	Flipkart	7500	E-Commerce
0	Paytm	5600	Fintech
4	Byjus	5400	EdTech
3	Zomato	4200	FoodTech
1	Ola	3800	Transport
2	Swiggy	3500	FoodTech
9	Delhivery	3000	Logistics
6	PhonePe	2800	Fintech
10	Meesho	2600	E-Commerce
7	Nykaa	2500	E-Commerce

/tmp/ipython-input-2690075798.py:31: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x='Investment_USD_Million', y='Startup', data=top10, palette='viridis')
```





## **Result:**

The Linear Regression model successfully analyzed startup investment data and identified the top-funded companies.