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## Machine Learning and Data Science Module 03 Exam

**Questions:** Suppose you are working on a data science project and have been given a dataset in CSV format called "sales\_data.csv". Your task is to analyze the data and extract meaningful insights using Python. Assume that you have already imported the necessary libraries and loaded the dataset into a Pandas DataFrame named "df".

### **Solution:**

## # Import necessary libraries

import pandas as pd
import matplotlib.pyplot as plt
df = pd.read csv("sales data.csv")

## I. Display the first 5 rows of the data frame.

print("I. Display the first 5 rows of the data frame:")
print(df.head())

#### II. Calculate the total number of rows in the DataFrame.

total\_rows = len(df)
print("\nII. Total number of rows in the DataFrame:", total rows)

## III. Check if there are any missing values in the dataset.

missing\_values = df.isnull().sum().any()
print("\nIII. Any missing values in the dataset:", missing values)

## IV. Calculate the average value of the "sales" column.

average\_sales = df["sales"].mean()
print("\nIV. Average value of the 'sales' column:", average sales)

# V. Create a new column called "profit" that contains the difference between the "sales" and "expenses" columns.

df["profit"] = df["sales"] - df["expenses"]

## VI. Sort the DataFrame in descending order based on the "profit" column.

df = df.sort\_values(by="profit", ascending=False)
print("\nVI. DataFrame sorted in descending order based on the 'profit' column:")

## VII. Save the sorted DataFrame to a new CSV file called "sorted\_sales\_data.csv".

df.to\_csv("sorted\_sales\_data.csv", index=False)

## VIII. Visualize the distribution of the "sales" column using a histogram plot.

plt.figure(figsize=(10, 6))
plt.hist(df["sales"], bins=20, color='blue', alpha=0.7)
plt.title("Histogram of Sales")

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```
plt.xlabel("Sales")
plt.ylabel("Frequency")
plt.show()
```

## IX. Create a scatter plot to explore the relationship between the "sales" and "expenses" columns.

```
plt.figure(figsize=(10, 6))
plt.scatter(df["sales"], df["expenses"], color='green', alpha=0.7)
plt.title("Scatter Plot: Sales vs Expenses")
plt.xlabel("Sales")
plt.ylabel("Expenses")
plt.show()
```

## X. Calculate the correlation coefficient between the "sales" and "expenses" columns.

```
correlation_coefficient = df["sales"].corr(df["expenses"])
print("\nX. Correlation coefficient between 'sales' and 'expenses':", correlation coefficient)
```

## XI. Calculate the total sales for each month and display the result.

```
monthly_total_sales = df.groupby("month")["sales"].sum()
print("\nXI. Total sales for each month:")
print(monthly_total_sales)
```

## XII. Find the maximum and minimum sales values for each year in the dataset.

```
yearly_max_sales = df.groupby("year")["sales"].max()
yearly_min_sales = df.groupby("year")["sales"].min()
print("\nXII. Maximum sales for each year:")
print(yearly_max_sales)
print("\n Minimum sales for each year:")
print(yearly_min_sales)
```

## XIII. Calculate the average sales for each category of products.

```
average_sales_by_category = df.groupby("category")["sales"].mean()
print("\nXIII. Average sales for each category of products:")
print(average_sales_by_category)
```

## XIV. Group the data by the "region" column and calculate the total sales for each region.

```
total_sales_by_region = df.groupby("region")["sales"].sum()
print("\nXIV. Total sales for each region:")
print(total_sales_by_region)
```

### XV. Calculate the sum of expenses for each quarter of the year.

```
expenses_by_quarter = df.groupby("quarter")["expenses"].sum()
print("\nXV. Sum of expenses for each quarter of the year:")
print(expenses_by_quarter)
```