

Exploratory Data Analysis (EDA) with Pandas in Retail Store

The purpose of this project is to explore and analyse the Retail Store dataset using the Pandas framework to derive insights into customer behaviour, product trends, and sales performance.

The goal of this analysis is to understand the structure, quality, and trends within the retail e-commerce data to make informed business decisions. The data contains information about products, customers, sales, and transactions.

Goals of the Project:

- Explore the Retail store dataset using Pandas.
- Perform feature engineering to derive useful insights.
- Visualize data distributions and trends with various plot types.
- Summarize key findings that can aid in business decision-making.

Materials and Methods

The data for this project is from a Retail store platform, containing information about orders, products, Payment Method, and Order quantity. This dataset includes sales data, product categories, order dates, Order quantity, and more. The analysis aims to understand sales performance, customer behaviour, product trends, and preferred shopping mode.

General Part

- **Libraries Import:** Pandas, NumPy, Seaborn, Matplotlib
- **Dataset Exploration:** Initial exploration of the dataset, checking for missing values, duplicates, and generating summary statistics.
- **Feature Engineering:** Transformation of date columns and creation of new features like shipping delay and profit margin.
- **Visualization in Pandas:** Distribution analysis, relationships between variables, and time-based trends.

Project Outcome & Insights

The project performs **Exploratory Data Analysis (EDA)** on an **Retail store Dataset** to gain meaningful insights into **sales performance, customer behavior, and Payment Methods**. Below are the key outcomes:

1. Sales Performance

- **Sales by Product Category:** The project groups sales based on different product category to identify the most popular products in the retail store.
- **Time Series Analysis:** It shows **sales trends over time**, helping businesses identify seasonal fluctuations and peak sales periods.
- **Top Performing Categories:** Identifies the product categories with the highest sales and revenue.

2. Customer Behaviour Analysis

- **Returning Customers:** The analysis helps in understanding customer retention by identifying customers who have made multiple purchases.
- **Top 10 High-Spending Customers:** Helps businesses recognize their most valuable customers and plan targeted marketing strategies.

3. Profitability & Business Growth

- **Top Performing Categories:** Helps understand **Top performing Categories** and identify areas for improving Product Availability and quality
- **Year-over-Year Sales Growth:** Tracks annual sales growth percentages, enabling better financial planning.

Feature Engineering:

Created new columns such as:

- **order_year, order_month, order_weekday** (Extracted from order_date).
- **returning_customer** (Boolean flag indicating repeated customers).

Key Questions and Insights to be Addressed:

- What is the total sales by Category?

```
sales_by_category =  
df.groupby('Category')['Quantity'].sum().sort_values(ascending=False)  
print("\nSales by Category:\n", sales_by_category)
```

- Which product categories have the highest sales?

```
sales_by_category =  
df.groupby('category_name')['sales_per_order'].sum().s  
ort_values(ascending=False)
```

Answer : Sales by Category:

Sales by Category:

Category

Food	7449.042839
Milk Products	7402.641686
Beverages	7383.509890
Furniture	7368.443992
Butchers	7355.042839
Computers and electric accessories	7233.641686
Electric household essentials	7179.108737
Patisserie	7139.504125

Name: Quantity, dtype: float64

Money Collected by a product Categorys:

Category	
Butchers	186374.684972
Electric household essentials	175502.836100
Beverages	175168.359408
Food	172468.684972
Furniture	170208.208279
Computers and electric accessories	165179.161664
Patisserie	164055.242869
Milk Products	158660.661664

- How does the sales trend change over time?

```
monthly_sales = df.groupby(['order_year',  
'order_month'])['Quantity'].sum().reset_index()  
  
print("Monthly Sales Trend:\n", monthly_sales)
```

Answer:

```
. Year-over-Year Sales Growth:  
  
order_year  
2022      NaN  
2023   -0.030834  
2024    0.085171  
2025   -0.950822
```

Name: Total Spent, dtype: float64

Monthly Sales Trend:

	order_year	order_month	Quantity
0	2022	2022-01	1896.191929
1	2022	2022-02	1582.461286
2	2022	2022-03	1591.862439
3	2022	2022-04	1520.263592
4	2022	2022-05	1469.461286
5	2022	2022-06	1577.928337
6	2022	2022-07	1699.191929
7	2022	2022-08	1693.395388
8	2022	2022-09	1543.263592
9	2022	2022-10	1482.461286
10	2022	2022-11	1628.191929
11	2022	2022-12	1504.658980
12	2023	2023-01	1870.796541
13	2023	2023-02	1443.395388
14	2023	2023-03	1521.994235
15	2023	2023-04	1424.461286
16	2023	2023-05	1607.994235
17	2023	2023-06	1642.928337
18	2023	2023-07	1687.527184
19	2023	2023-08	1345.796541
20	2023	2023-09	1499.395388
21	2023	2023-10	1492.928337

22	2023	2023-11	1372.862439
23	2023	2023-12	1590.060133
24	2024	2024-01	1835.658980
25	2024	2024-02	1472.928337
26	2024	2024-03	1619.593082
27	2024	2024-04	1682.461286
28	2024	2024-05	1698.593082
29	2024	2024-06	1658.461286
30	2024	2024-07	1602.197694
31	2024	2024-08	1695.527184
32	2024	2024-09	1490.329490
33	2024	2024-10	1580.395388
34	2024	2024-11	1603.395388
35	2024	2024-12	1901.126031
36	2025	2025-01	980.796541

- What is the preferred Shopping mode?

```
sales_by_location =
df.groupby('Location')['Quantity'].sum().sort_values(a
scending=False)

print("\nSales by location:\n", sales_by_location)
```

Answer:

Sales by location:

Location

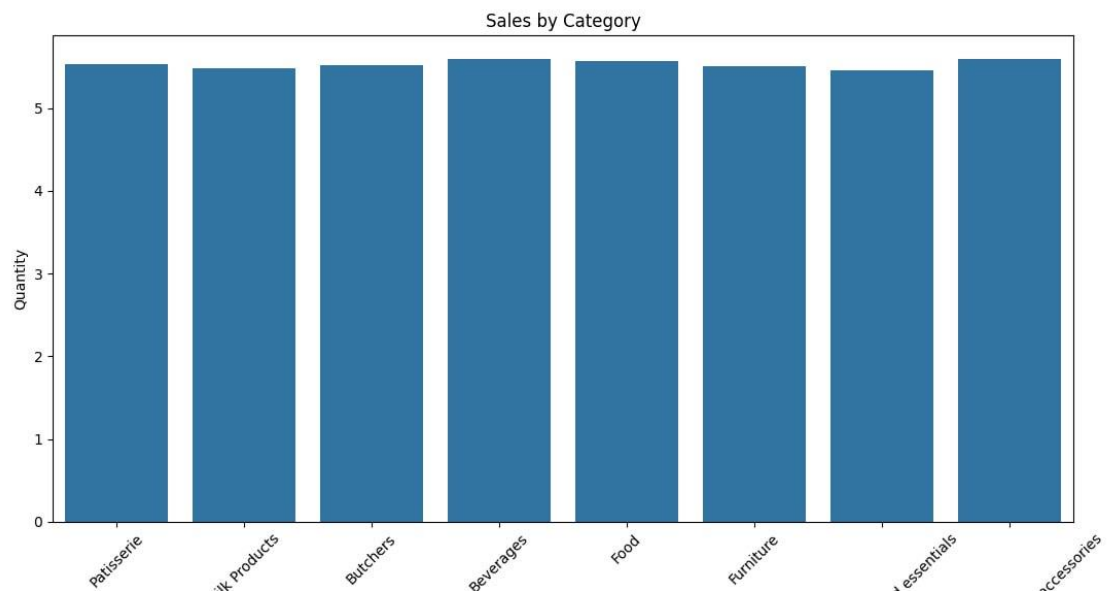
In-store 34304.477786

Online 24206.458006

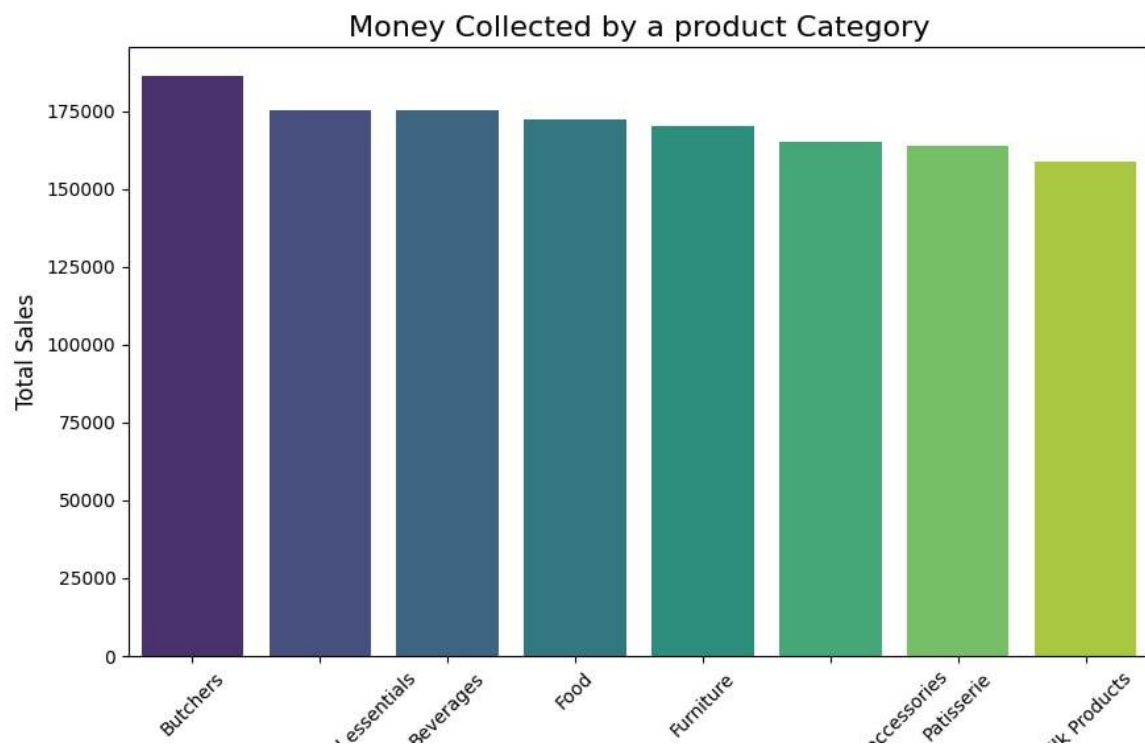
Visualization:

Several charts created to present inside including:

- Sales by Categoryn (Bar chart)



- Money collecetby product category (Bar chart)



- Sales trends over time (Line chart)

