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## 0.1 DATA ANALYSIS PYTHON PROJECT - BLINKIT ANALYSIS

### 0.1.1 Import Libraries

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

### 0.1.2 Import Raw data

```
[2]: df = pd.read_csv("C:/Blinkit-Analysis/blinkit_data.csv")
```

### 0.1.3 Sample data

```
[3]: df.head(10)
```

```
[3]:  Item Fat Content Item Identifier      Item Type \
0      Regular      FDX32  Fruits and Vegetables
1      Low Fat      NCB42    Health and Hygiene
2      Regular      FDR28    Frozen Foods
3      Regular      FDL50          Canned
4      Low Fat      DRI25    Soft Drinks
5      low fat      FDS52    Frozen Foods
6      Low Fat      NCU05    Health and Hygiene
7      Low Fat      NCD30    Household
8      Low Fat      FDW20  Fruits and Vegetables
9      Low Fat      FDX25          Canned

      Outlet Establishment Year Outlet Identifier Outlet Location Type \
0              2012      OUT049          Tier 1
1              2022      OUT018          Tier 3
2              2010      OUT046          Tier 1
3              2000      OUT013          Tier 3
4              2015      OUT045          Tier 2
5              2020      OUT017          Tier 2
6              2011      OUT010          Tier 3
7              2015      OUT045          Tier 2
```

8	2000	OUT013	Tier 3
9	1998	OUT027	Tier 3

	Outlet Size	Outlet Type	Item Visibility	Item Weight	Sales \
0	Medium	Supermarket Type1	0.100014	15.10	145.4786
1	Medium	Supermarket Type2	0.008596	11.80	115.3492
2	Small	Supermarket Type1	0.025896	13.85	165.0210
3	High	Supermarket Type1	0.042278	12.15	126.5046
4	Small	Supermarket Type1	0.033970	19.60	55.1614
5	Small	Supermarket Type1	0.005505	8.89	102.4016
6	Small	Grocery Store	0.098312	11.80	81.4618
7	Small	Supermarket Type1	0.026904	19.70	96.0726
8	High	Supermarket Type1	0.024129	20.75	124.1730
9	Medium	Supermarket Type3	0.101562	NaN	181.9292

	Rating
0	5.0
1	5.0
2	5.0
3	5.0
4	5.0
5	5.0
6	5.0
7	5.0
8	5.0
9	5.0

```
[4]: df.tail(10)
```

	Item Fat Content	Item Identifier	Item Type \
8513	Regular	DRY23	Soft Drinks
8514	low fat	FDA11	Baking Goods
8515	low fat	FDK38	Canned
8516	low fat	FD038	Canned
8517	low fat	FDG32	Fruits and Vegetables
8518	low fat	NCT53	Health and Hygiene
8519	low fat	FDN09	Snack Foods
8520	low fat	DRE13	Soft Drinks
8521	reg	FDT50	Dairy
8522	reg	FDM58	Snack Foods

	Outlet Establishment	Year	Outlet Identifier	Outlet Location Type \
8513		1998	OUT027	Tier 3
8514		1998	OUT027	Tier 3
8515		1998	OUT027	Tier 3
8516		1998	OUT027	Tier 3
8517		1998	OUT027	Tier 3

8518	1998	OUT027	Tier 3
8519	1998	OUT027	Tier 3
8520	1998	OUT027	Tier 3
8521	1998	OUT027	Tier 3
8522	1998	OUT027	Tier 3

	Outlet Size	Outlet Type	Item Visibility	Item Weight	Sales \
8513	Medium	Supermarket Type3	0.108568	NaN	42.9112
8514	Medium	Supermarket Type3	0.043029	NaN	94.7436
8515	Medium	Supermarket Type3	0.053032	NaN	149.1734
8516	Medium	Supermarket Type3	0.072486	NaN	78.9986
8517	Medium	Supermarket Type3	0.175143	NaN	222.3772
8518	Medium	Supermarket Type3	0.000000	NaN	164.5526
8519	Medium	Supermarket Type3	0.034706	NaN	241.6828
8520	Medium	Supermarket Type3	0.027571	NaN	86.6198
8521	Medium	Supermarket Type3	0.107715	NaN	97.8752
8522	Medium	Supermarket Type3	0.000000	NaN	112.2544

	Rating
8513	4.0
8514	4.0
8515	4.0
8516	4.0
8517	4.0
8518	4.0
8519	4.0
8520	4.0
8521	4.0
8522	4.0

#### 0.1.4 Size of Data

```
[5]: print(" Size of data:",df.shape)
```

```
Size of data: (8523, 12)
```

#### 0.1.5 Field Info

```
[6]: df.columns
```

```
[6]: Index(['Item Fat Content', 'Item Identifier', 'Item Type',
         'Outlet Establishment Year', 'Outlet Identifier',
         'Outlet Location Type', 'Outlet Size', 'Outlet Type', 'Item Visibility',
         'Item Weight', 'Sales', 'Rating'],
         dtype='object')
```

### 0.1.6 Data Types

```
[7]: df.dtypes
```

```
[7]: Item Fat Content      object
     Item Identifier      object
     Item Type            object
     Outlet Establishment Year  int64
     Outlet Identifier      object
     Outlet Location Type    object
     Outlet Size           object
     Outlet Type           object
     Item Visibility        float64
     Item Weight           float64
     Sales                 float64
     Rating               float64
     dtype: object
```

### 0.1.7 Data Cleaning

```
[8]: print(df['Item Fat Content'].unique())
```

```
['Regular' 'Low Fat' 'low fat' 'LF' 'reg']
```

```
[9]: df['Item Fat Content'] = df['Item Fat Content'].replace({'LF': 'Low Fat',
                                                             'low fat' : 'Low Fat',
                                                             'reg' : 'Low Fat'})
```

```
[10]: print(df['Item Fat Content'].unique())
```

```
['Regular' 'Low Fat']
```

### 0.1.8 BUSINESS REQUIREMENTS

#### 0.1.9 KPI'S REQUIREMENTS

```
[11]: # Total Sales
total_sales = df['Sales'].sum()

# Average Sales
avg_sales = df['Sales'].mean()

#No of Items Sold
no_of_items_sold = df['Sales'].count()

#Average Ratings
avg_ratings = df['Rating'].mean()
```

```
#Display
print(f"Total Sales: ${total_sales:,.1f}")
print(f"Total Sales: ${avg_sales:,.1f}")
print(f"Total Sales: ${no_of_items_sold:,.1f}")
print(f"Total Sales: ${avg_ratings:,.1f}")
```

```
Total Sales: $1,201,681.5
Total Sales: $141.0
Total Sales: $8,523.0
Total Sales: $4.0
```

### 0.1.10 CHARTS REQUIREMENTS

#### 0.1.11 Total Sales by Fat Content

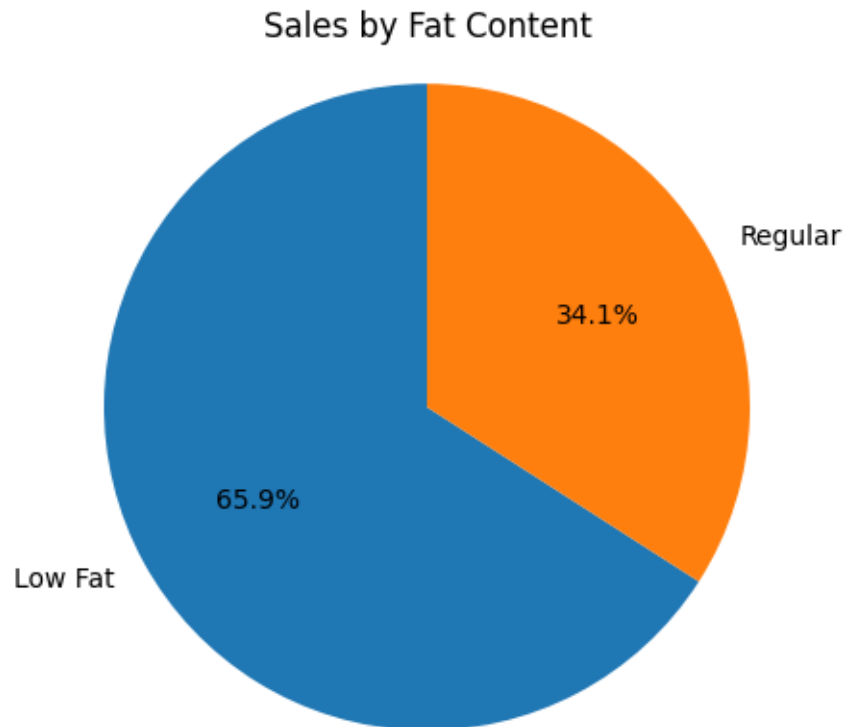
```
[12]: sales_by_fat = df.groupby('Item Fat Content')['Sales'].sum()

plt.pie(sales_by_fat, labels= sales_by_fat.index,
        autopct = '%.1f%%',
        startangle = 90)

plt.title('Sales by Fat Content')

plt.axis('equal')

plt.show()
```



#### 0.1.12 Total Sales by Item Type

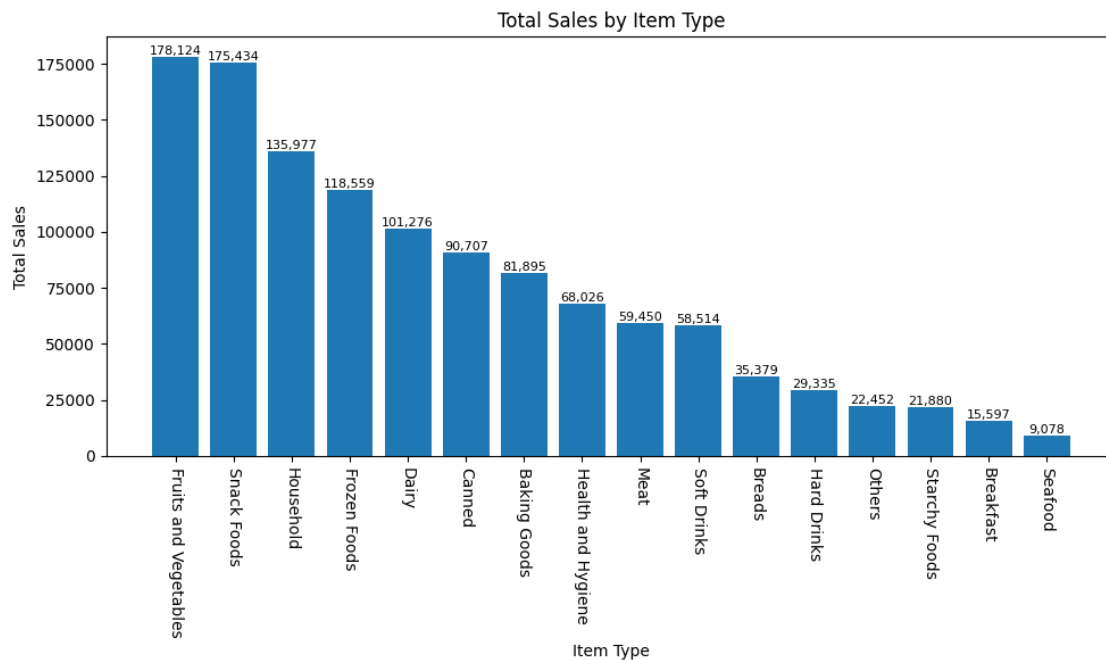
```
[13]: sales_by_type = df.groupby('Item Type')['Sales'].sum().
      ↪sort_values(ascending=False)

plt.figure(figsize=(10, 6))
bars = plt.bar(sales_by_type.index, sales_by_type.values)

plt.xticks(rotation=-90)
plt.xlabel('Item Type')
plt.ylabel('Total Sales')
plt.title('Total Sales by Item Type')

for bar in bars:
    plt.text(
        bar.get_x() + bar.get_width() / 2,
        bar.get_height(),
        f'{bar.get_height():,.0f}',
        ha='center',
        va='bottom',
        fontsize=8
    )
```

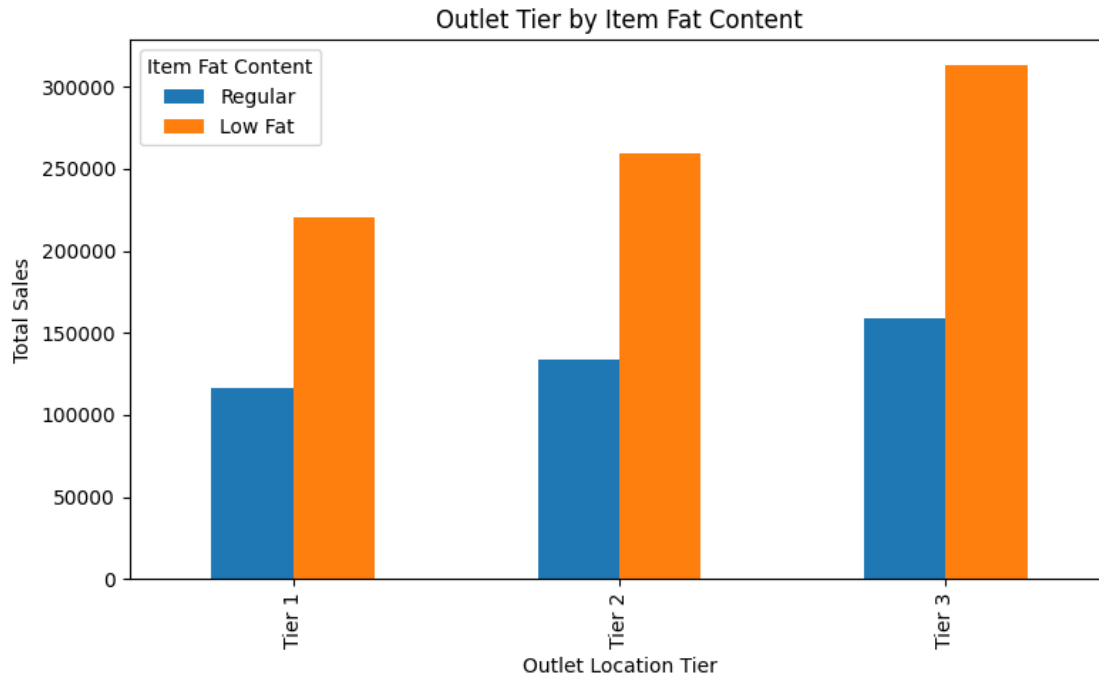
```
plt.tight_layout()
plt.show()
```



### 0.1.13 Fat Content by Outlet for Total Sales

```
[16]: grouped = df.groupby(['Outlet Location Type', 'Item Fat Content'])['Sales'].
      ↪sum().unstack()
grouped = grouped[['Regular', 'Low Fat']]

ax = grouped.plot(kind='bar', figsize=(8, 5), title='Outlet Tier by Item Fat_
      ↪Content')
plt.xlabel('Outlet Location Tier')
plt.ylabel('Total Sales')
plt.legend(title='Item Fat Content')
plt.tight_layout()
plt.show()
```



#### 0.1.14 Total Sales by Outlet Establishment

```
[17]: sales_by_year = df.groupby('Outlet Establishment Year')['Sales'].sum().
      ↪sort_index()

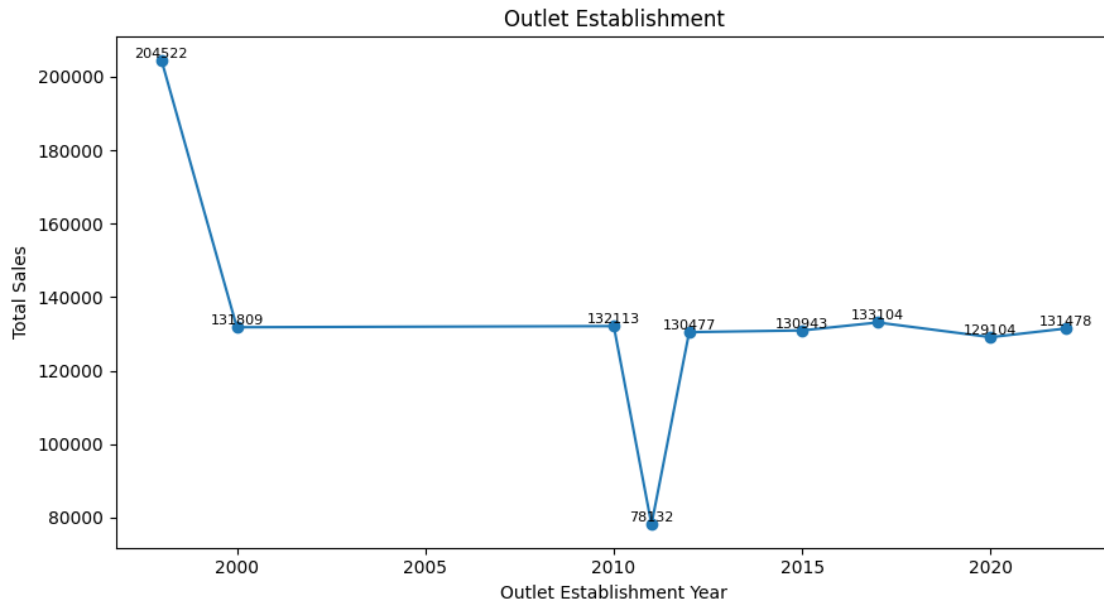
plt.figure(figsize=(9,5))
plt.plot(sales_by_year.index, sales_by_year.values, marker='o', linestyle='-')

plt.xlabel('Outlet Establishment Year')
plt.ylabel('Total Sales')
plt.title('Outlet Establishment')

for x, y in zip(sales_by_year.index, sales_by_year.values):
    plt.text(x, y, f'{y:.0f}', ha='center', va='bottom', fontsize=8)

plt.tight_layout()
plt.show()
```

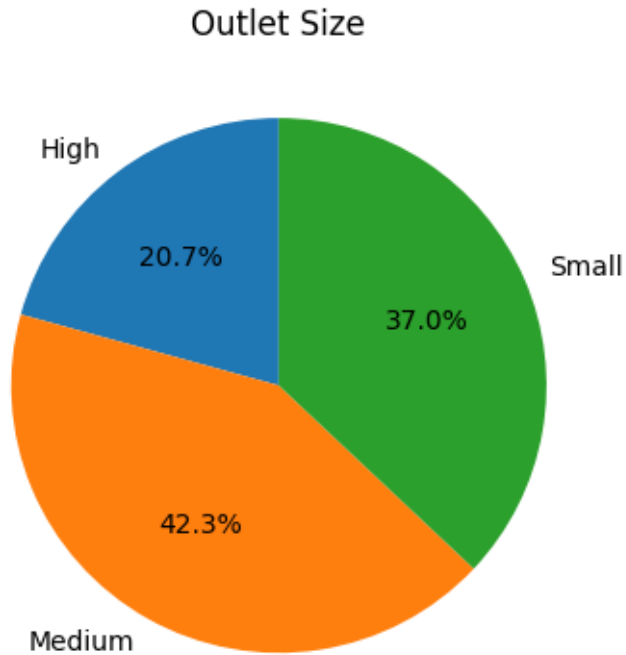




### 0.1.15 Sales by Outlet Size

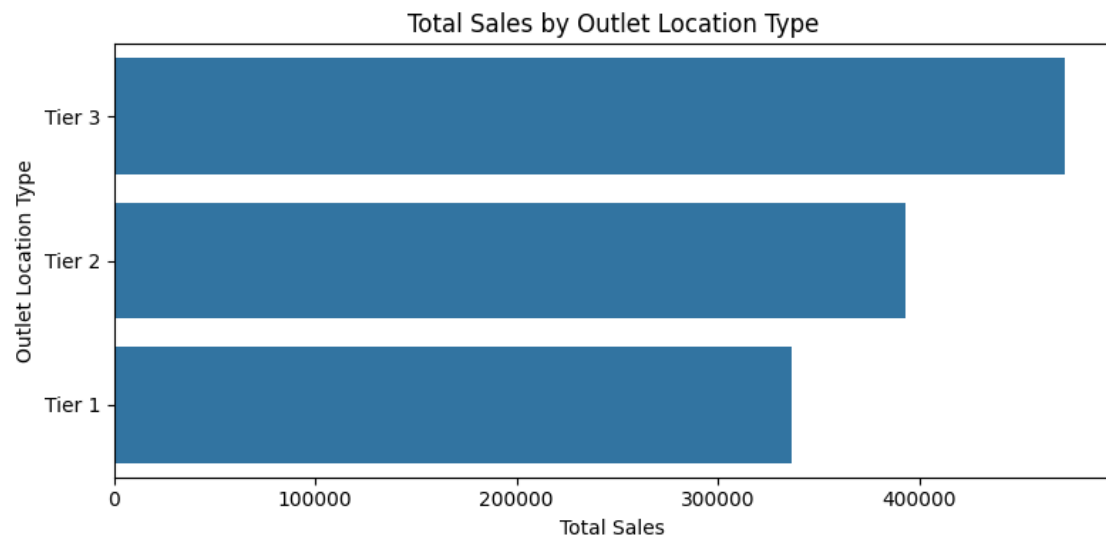
```
[18]: sales_by_size = df.groupby('Outlet Size')['Sales'].sum()

plt.figure(figsize=(4, 4))
plt.pie(sales_by_size, labels=sales_by_size.index, autopct='%1.1f%%',
        ↪startangle=90)
plt.title('Outlet Size')
plt.tight_layout()
plt.show()
```



#### 0.1.16 Sales by Outlet Location

```
[20]: sales_by_location = df.groupby('Outlet Location Type')['Sales'].sum().  
      ↪reset_index()  
sales_by_location = sales_by_location.sort_values('Sales', ascending=False)  
  
plt.figure(figsize=(8, 4)) # Smaller height, enough width  
ax = sns.barplot(x='Sales', y='Outlet Location Type', data=sales_by_location)  
  
plt.title('Total Sales by Outlet Location Type')  
plt.xlabel('Total Sales')  
plt.ylabel('Outlet Location Type')  
  
plt.tight_layout() # Ensures layout fits without scroll  
plt.show()
```



[ ]: