

# Blinkit Analysis

February 23, 2026

## 1 Data Analysis Python Project

### 1.1 Import Libraries

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

### 1.2 Importing Raw Data

```
[3]: df = pd.read_csv("blinkit_data.csv")
```

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

```
[4]:   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

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

Size of Data: (8523, 12)

```
[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')
```

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

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
```

```
[9]: df.describe()
```

```
[9]:    Outlet Establishment Year  Item Visibility  Item Weight      Sales  \\\n  count           8523.000000   8523.000000  7060.000000  8523.000000  
  mean            2010.831867   0.066132    12.857645   140.992782  
  std             8.371760    0.051598    4.643456    62.275067  
  min             1998.000000   0.000000    4.555000    31.290000  
  25%            2000.000000   0.026989    8.773750    93.826500  
  50%            2012.000000   0.053931   12.600000   143.012800  
  75%            2017.000000   0.094585   16.850000   185.643700  
  max            2022.000000   0.328391   21.350000   266.888400  
  
  Rating  
  count  8523.000000  
  mean    3.965857  
  std     0.605651  
  min     1.000000  
  25%    4.000000  
  50%    4.000000  
  75%    4.200000  
  max    5.000000
```

### 1.3 Data Cleaning

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

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

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

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

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

### 1.4 KPI's Requirement

```
[15]: # Total Sales\ntotal_sales = df['Sales'].sum()\n\n# Average Sales\navg_sales = df['Sales'].mean()
```

```

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

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

# Display
print(f'Total Sales: ${total_sales:,.0f}')
print(f'Average Sales: ${avg_sales:,.0f}')
print(f'No of Items Sold: {no_of_items_sold:,.0f}')
print(f'Average Rating: {avg_ratings:.1f}')

```

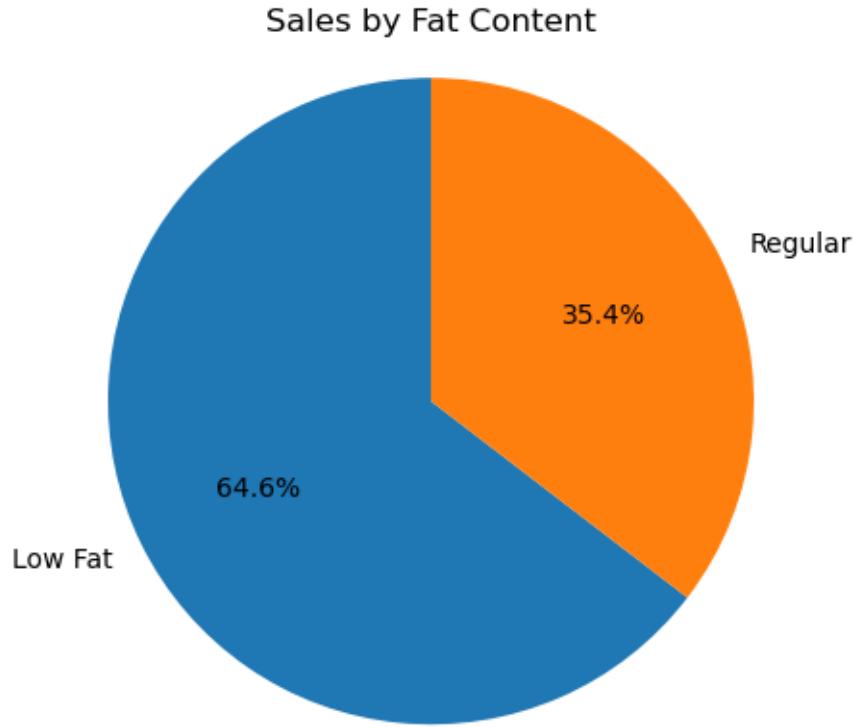
Total Sales: \$ 1,201,681  
 Average Sales: \$ 141  
 No of Items Sold: 8,523  
 Average Rating: 4.0

## 1.5 Charts Requirement

### 1.5.1 Total Sales by Fat Content

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

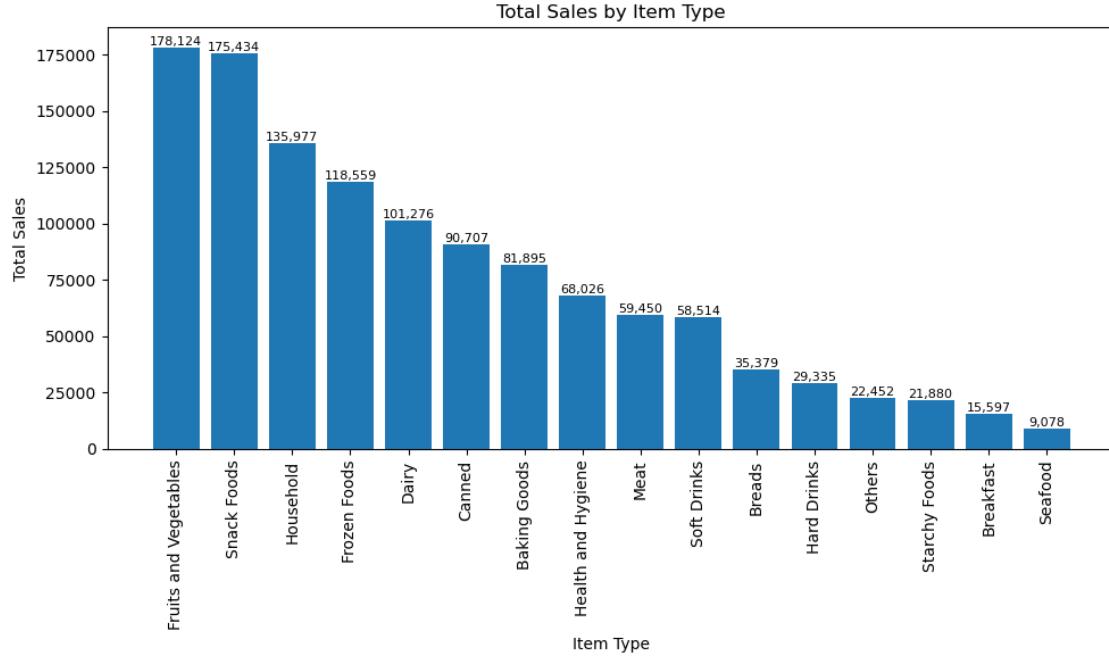
plt.pie(sales_by_fat, labels = sales_by_fat.index,
         autopct = '%.1f%%',
         startangle= 90)
plt.title('Sales by Fat Content')
plt.axis('equal')
plt.show()
```



#### 1.5.2 Total Sales by Item Type

```
[23]: 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  
)
```

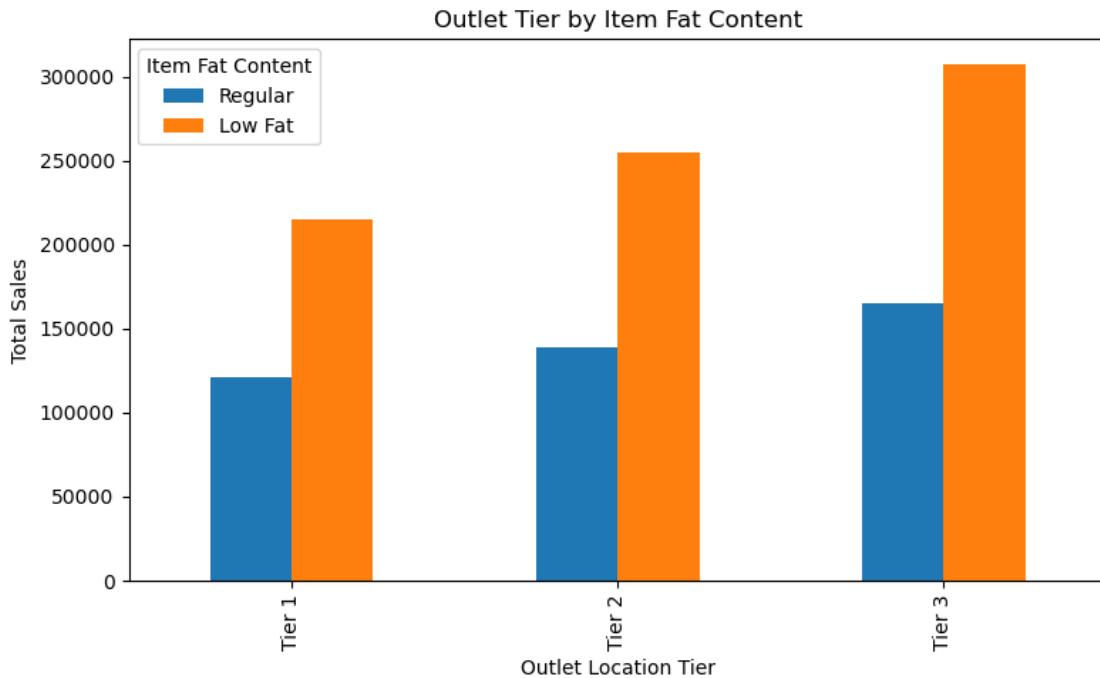
```
plt.tight_layout()
plt.show()
```



### 1.5.3 Fat Content by Outlet for Total Sales

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

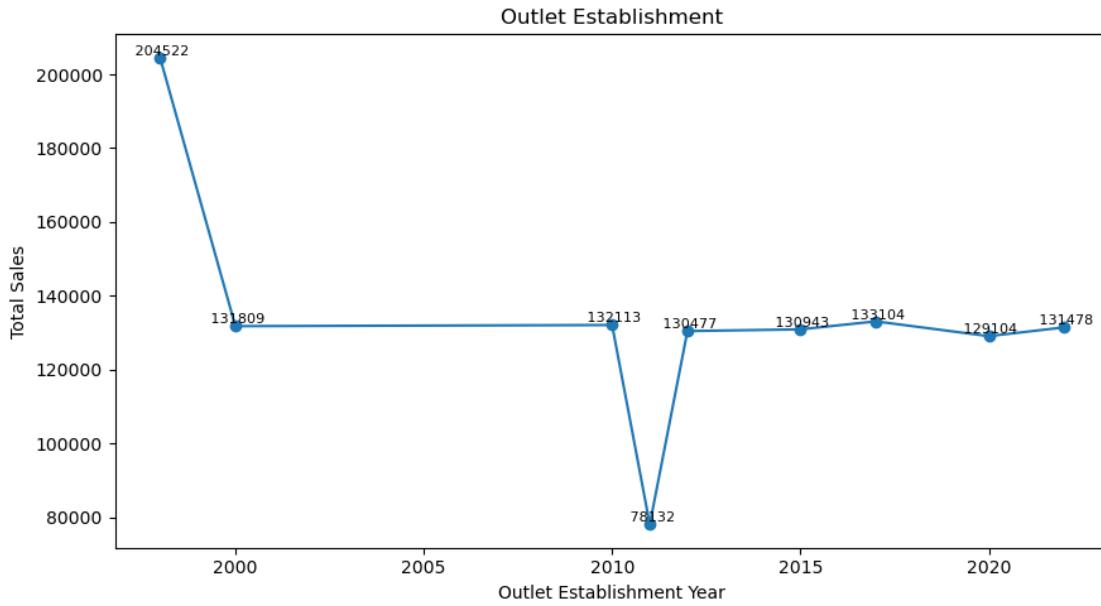
ax = grouped.plot(kind="bar", figsize=(8, 5), title = 'Outlet Tier by Item Fat Content')
plt.xlabel('Outlet Location Tier')
plt.ylabel('Total Sales')
plt.legend(title='Item Fat Content')
plt.tight_layout()
plt.show()
```



#### 1.5.4 Total Sales by Outlet Establishment

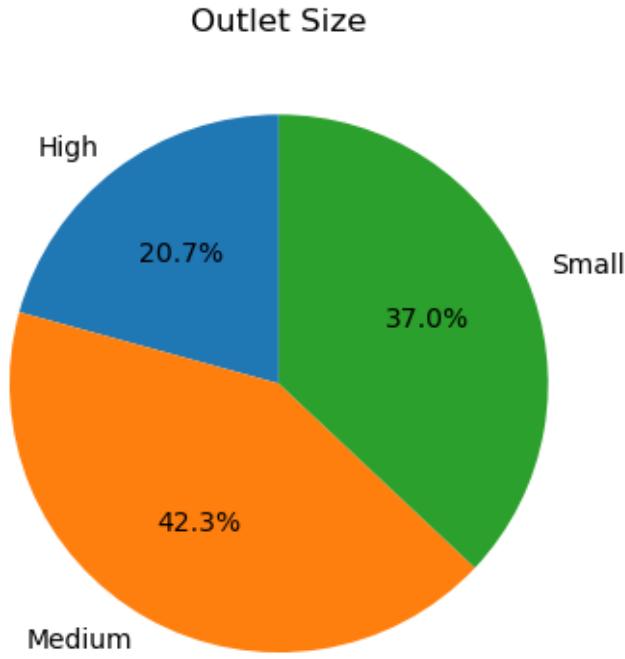
```
[27]: sales_by_year = df.groupby('Outlet Establishment Year')['Sales'].sum().
    sort_index()
plt.figure(figsize = (9,5))
plt.plot(sales_by_year.index, sales_by_year.values, marker= "o", linestyle = "-")
plt.xlabel("Outlet Establishment Year")
plt.ylabel('Total Sales')
plt.title('Outlet Establishment')

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()
```



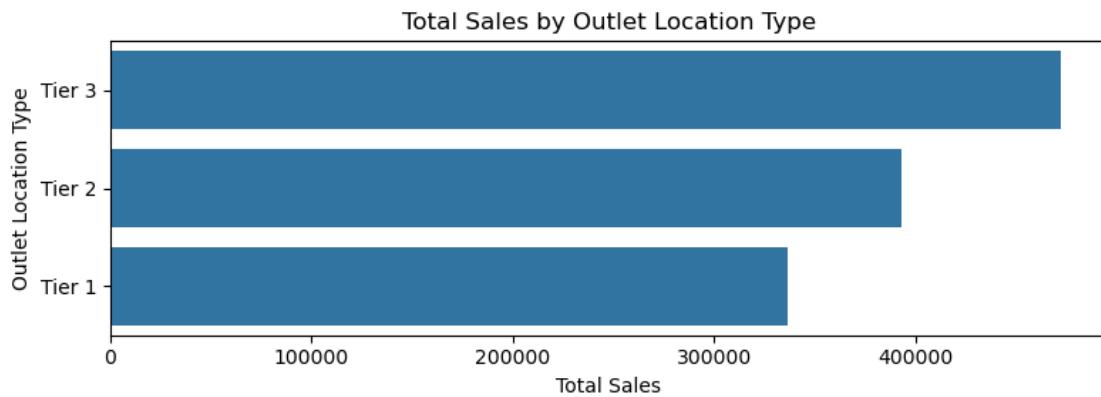
### 1.5.5 Sales by Outlet Size

```
[28]: 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='%.1f%%',
        startangle=90)
plt.title("Outlet Size")
plt.tight_layout()
plt.show()
```



#### 1.5.6 Sales by Outlet Location

```
[29]: 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, 3))  
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()  
plt.show()
```



[ ]: