

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
```

```
[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

```

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

```

[9]:      Outlet Establishment Year  Item Visibility  Item Weight      Sales \
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',
                                                                'low fat': 'Low Fat',
                                                                'reg': 'Regular'})
```

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

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

1.4 KPI's Requirement

```

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

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

```

```

# No. of Itmes 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

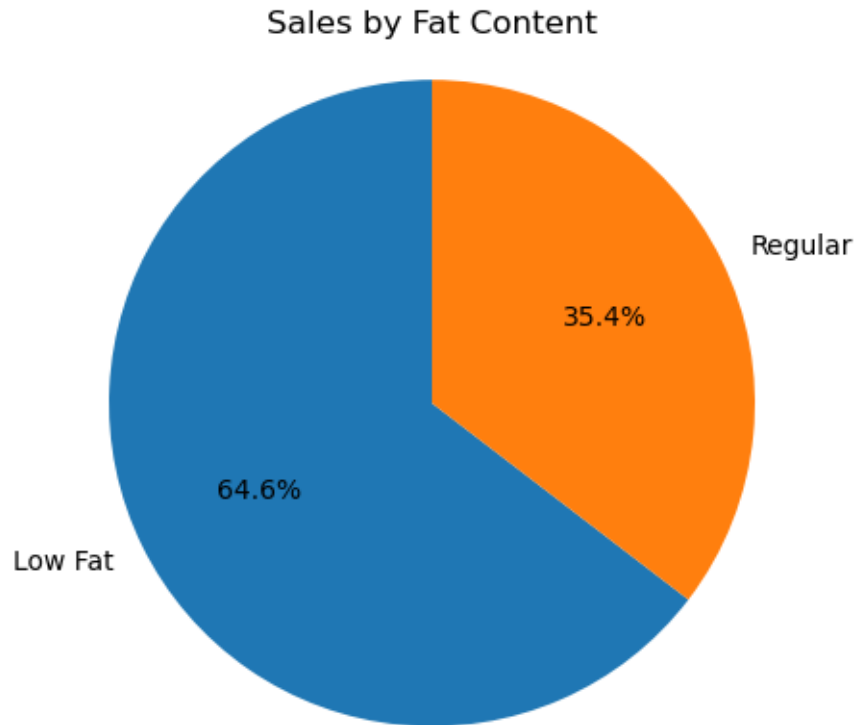
1.5.1 Total Sales by Fat Content

```

[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()

```



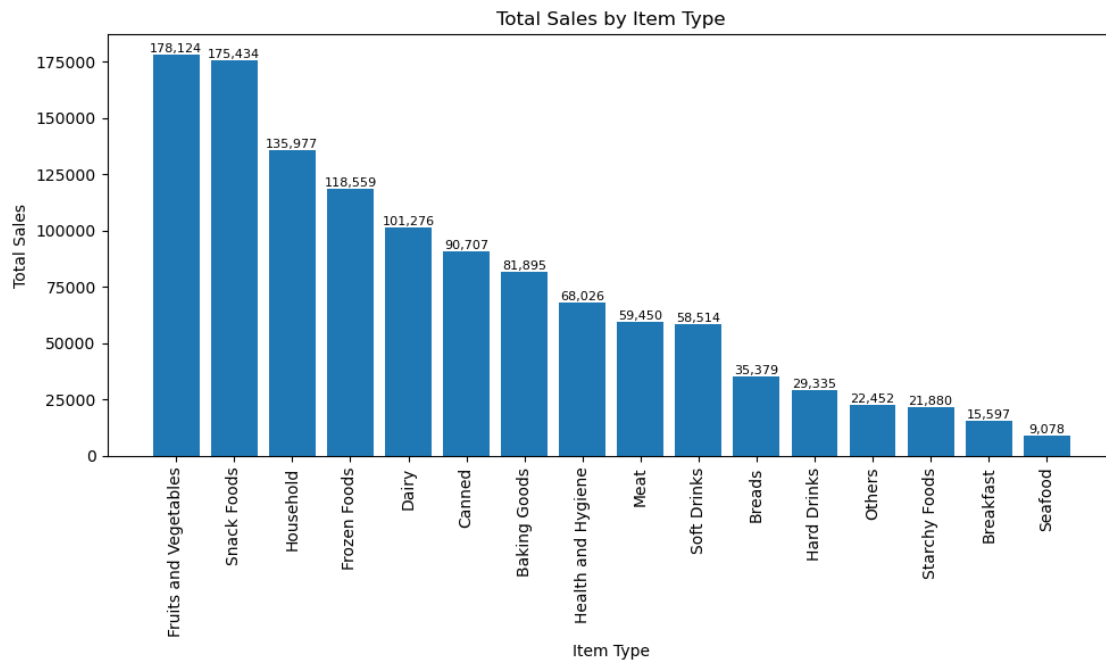
1.5.2 Total Sales by Item Type

```
[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
    )
```

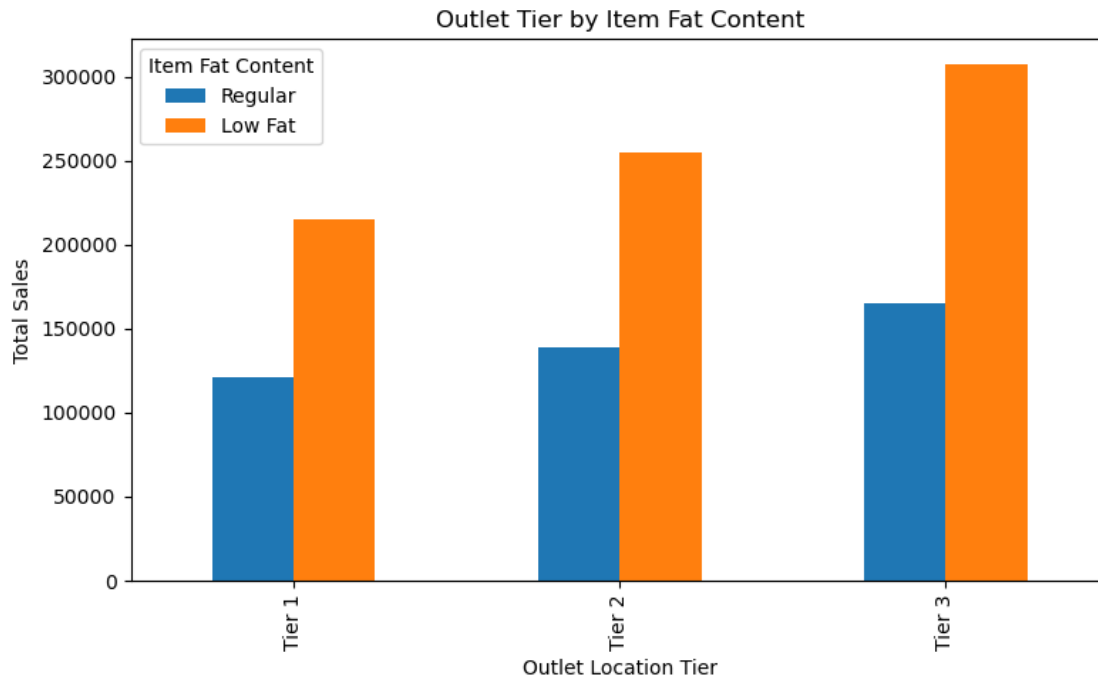
```
plt.tight_layout()
plt.show()
```



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']]

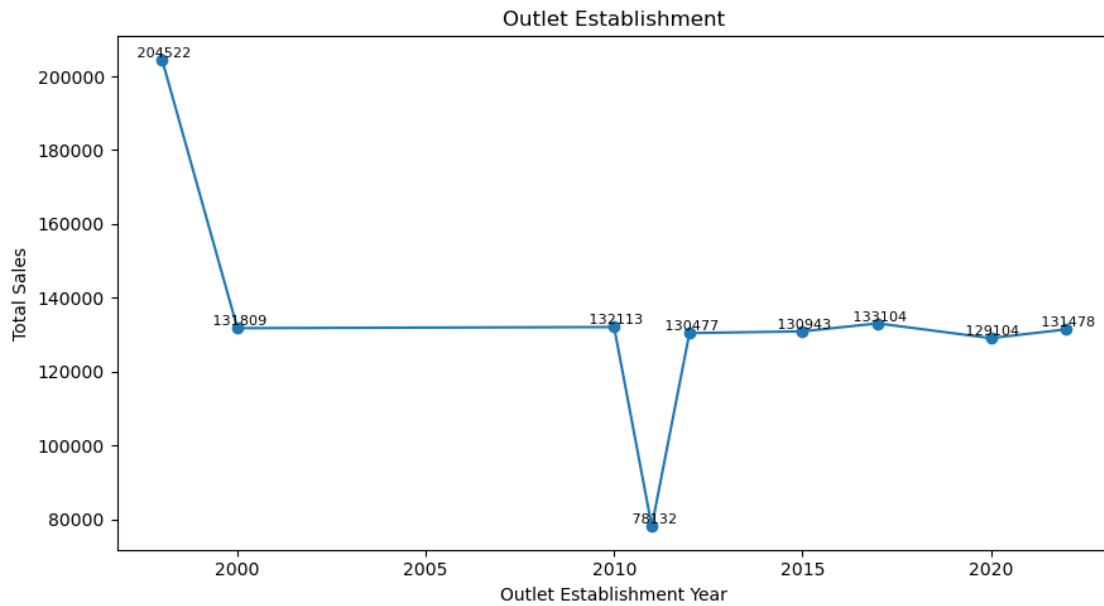
ax = grouped.plot(kind="bar", figsize=(8, 5), title = 'Outlet Tier by Item Fat_
      ↪Content')
plt.xlabel('Outlet Location Tier')
plt.ylabel('Total Sales')
plt.legend(title='Item Fat Content')
plt.tight_layout()
plt.show()
```



1.5.4 Total Sales by Outlet Establishment

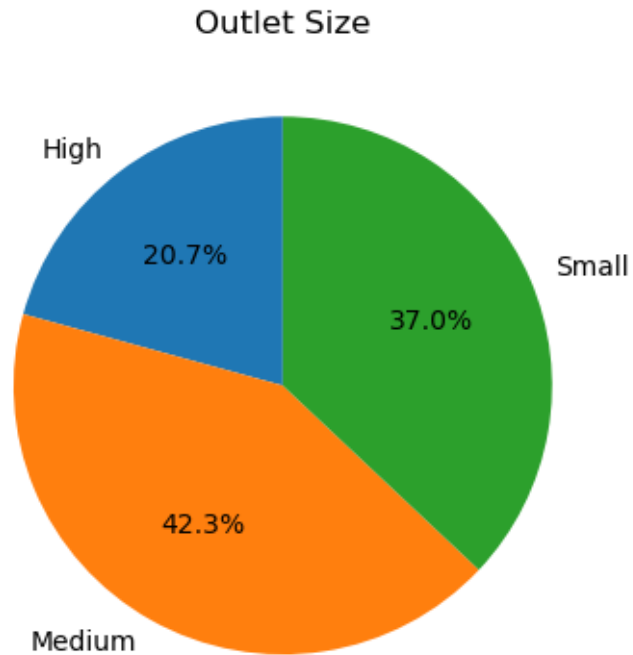
```
[27]: sales_by_year = df.groupby('Outlet Establishment Year')['Sales'].sum().
      ↪sort_index()
plt.figure(figsize = (9,5))
plt.plot(sales_by_year.index, sales_by_year.values, marker= "o", linestyle =_
      ↪"-")
plt.xlabel("Outlet Establishment Year")
plt.ylabel('Total Sales')
plt.title('Outlet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



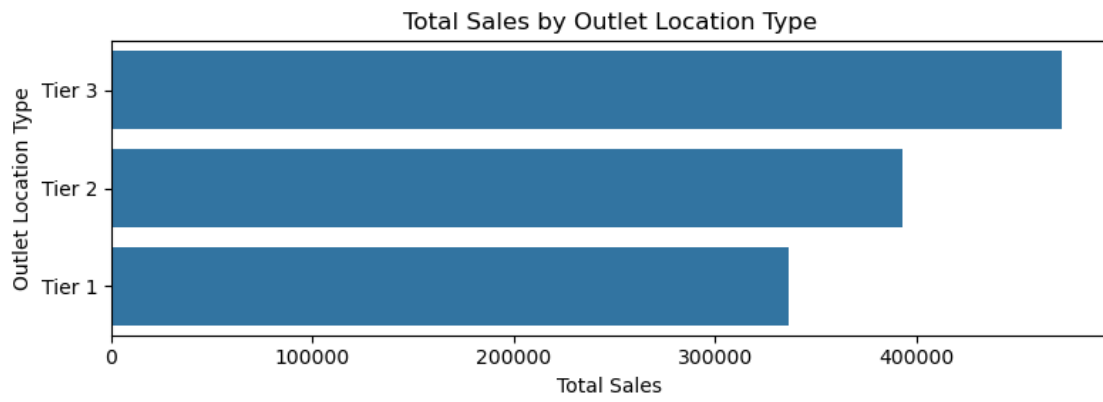
1.5.5 Sales by Outlet Size

```
[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='%1.1f%%',
        ↪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()
```



[]: