Grocery store sales analysis

August 27, 2024

```
[1]: pip install --upgrade pandas numpy openpyxl
     pip install --upgrade numpy statsmodels
    Requirement already up-to-date: pandas in ./opt/anaconda3/lib/python3.8/site-
    packages (2.0.3)
    Requirement already up-to-date: numpy in ./opt/anaconda3/lib/python3.8/site-
    packages (1.24.4)
    Requirement already up-to-date: openpyxl in ./opt/anaconda3/lib/python3.8/site-
    packages (3.1.5)
    Requirement already satisfied, skipping upgrade: tzdata>=2022.1 in
    ./opt/anaconda3/lib/python3.8/site-packages (from pandas) (2023.4)
    Requirement already satisfied, skipping upgrade: python-dateutil>=2.8.2 in
    ./opt/anaconda3/lib/python3.8/site-packages (from pandas) (2.9.0.post0)
    Requirement already satisfied, skipping upgrade: pytz>=2020.1 in
    ./opt/anaconda3/lib/python3.8/site-packages (from pandas) (2020.1)
    Requirement already satisfied, skipping upgrade: et-xmlfile in
    ./opt/anaconda3/lib/python3.8/site-packages (from openpyxl) (1.0.1)
    Requirement already satisfied, skipping upgrade: six>=1.5 in
    ./opt/anaconda3/lib/python3.8/site-packages (from python-
    dateutil>=2.8.2->pandas) (1.15.0)
    Note: you may need to restart the kernel to use updated packages.
[1]: import math
     import matplotlib.pyplot as plt
     import numpy as np
     import pandas as pd
     import random
     import re
     import scipy
     from scipy.stats import norm
     from scipy import stats as st
     import seaborn as sns
     import statsmodels.api as sm
```

1 Preliminary data analysis

```
[2]: path_to_file = 'Grocery Database.xlsx'
df = pd.read_excel(path_to_file)
```

Data shape overview

```
[18]: df.shape print(f'The set contains {df.shape[0]} rows and {df.shape[1]} columns.')
```

The set contains 50447 rows and 32 columns.

Let us review the data types in the dataset

```
[19]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50447 entries, 0 to 50446
Data columns (total 32 columns):

Data	Columns (Cotal S2 Columns).		
#	Column	Non-Null Count	Dtype
0	Receipt Number	50447 non-null	object
1	Date	50447 non-null	datetime64[ns]
2	Year	50447 non-null	int64
3	Month	50447 non-null	int64
4	Time	50447 non-null	object
5	Mobile Number	50447 non-null	int64
6	Membership ID	50447 non-null	object
7	Loyalty Card Points	50447 non-null	int64
8	Age	50447 non-null	int64
9	Gender	50447 non-null	object
10	City	50447 non-null	object
11	Country	50447 non-null	object
12	Category	50447 non-null	object
13	Sub_Category	50447 non-null	object
14	Items	50447 non-null	object
15	Brand	50447 non-null	object
16	Description	50447 non-null	object
17	Price	50447 non-null	float64
18	QTY	50447 non-null	int64
19	DISC	50447 non-null	float64
20	Amount	50447 non-null	float64
21	Net Bill Amount	50447 non-null	float64
22	GST	50447 non-null	float64
23	Gross Bill Amount	50447 non-null	float64
24	Payment Mode	50447 non-null	object
25	Bank Name	50447 non-null	object
26	% Profit Margin	50447 non-null	float64
27	% Operating Cost	50447 non-null	float64

```
28 % Product Cost 50447 non-null float64
29 Profit Margin 50447 non-null float64
30 Operating Cost 50447 non-null float64
31 Product Cost 50447 non-null float64
dtypes: datetime64[ns](1), float64(12), int64(6), object(13)
memory usage: 12.3+ MB
```

Let us check whether the data set has missing values

```
[20]: df.isna().sum()
[20]: Receipt Number
                               0
                               0
      Date
                               0
      Year
      Month
                               0
      Time
                               0
      Mobile Number
                               0
      Membership ID
                               0
      Loyalty Card Points
                               0
                               0
      Age
      Gender
                               0
      City
                               0
                               0
      Country
      Category
                               0
                               0
      Sub_Category
                               0
      Items
                               0
      Brand
                               0
      Description
      Price
                               0
      QTY
                               0
      DISC
                               0
      Amount
                               0
      Net Bill Amount
                               0
      GST
                               0
      Gross Bill Amount
                               0
      Payment Mode
                               0
      Bank Name
                               0
      % Profit Margin
                               0
      % Operating Cost
                               0
      % Product Cost
                               0
                               0
      Profit Margin
                               0
      Operating Cost
      Product Cost
                               0
      dtype: int64
```

There are no missing values in the data set.

2 Main part of data analysis

1: What is the most common way of payment among costumers?

```
[21]: df['Payment Mode'].value_counts()
```

[21]: Payment Mode

Card 17149 Wallet 16867 Cash 16431

Name: count, dtype: int64

Conclusion: We can see that the most popular way of payment among customesr is card.

Business idea: One can develop a mobile app that allows customers to make contactless payments. This could include features like order-ahead and curbside pickup, creating a convenient and safe shopping experience.

2: What are the most commonly purchased product categories?

```
[22]: df.rename(columns = {'Receipt Number':'Receipt_Number'}, inplace = True)
df.groupby(['Category']).Receipt_Number.count()
```

[22]: Category

Bakery & Breakfast 6770 Beauty 5205 Beverages 2208 Choco, Snacks, Sweets 3019 Dairy, Chilled & Eggs 3381 Frozen 6393 Fruit & Vegetable 4370 Health 1008 Household 1881 Kitchen & Dining 2552 Meat & Seafood 2149 Mother & Baby 2064 Party Supplies 1836 Pet Care 3465 Rice & Cooking Essentials 2432 Wines, Beers & Spirits 1714 Name: Receipt_Number, dtype: int64

Conclusion: One sees that the "Bakery and breakfast" is the most purchuased category.

Business ideas: 1. Gourmet Breakfast Delivery Service: Create a business that specializes in delivering gourmet breakfast options to people's homes or offices.

2.Healthy Breakfast Subscription Box: Develop a subscription box service that delivers healthy breakfast options to subscribers each month. Include items like granola, yogurt, whole-grain muffins, and fresh fruits.

3: What is the average price for the categories?

```
[15]: df.groupby(['Category']).Price.mean()
```

[15]: Category Bakery & Breakfast 5.222038 Beauty 10.063051 Beverages 8.297360 Choco, Snacks, Sweets 5.770414 Dairy, Chilled & Eggs 6.900683 Frozen 7.692615 Fruit & Vegetable 5.570590 Health 13.693046 Household 8.855045 Kitchen & Dining 3.753100 Meat & Seafood 11.947627 Mother & Baby 11.869695 Party Supplies 4.641950 Pet Care 6.976049 Rice & Cooking Essentials 3.507113 Wines, Beers & Spirits 13.282456

Name: Price, dtype: float64

Conclusion: The highest price is in the "Health" category.

Business idea: Beauty Tech: Invest in beauty technology, such as AI-powered skincare analysis or virtual makeup try-on apps. You could provide these services to consumers, helping them find the right products for their skin type and tone.

4: What are the top-selling products in terms of quantity sold or revenue generated?

```
[25]: df.rename(columns = {'Profit Margin':'Profit_Margin'}, inplace = True)
df.groupby('Items').Amount.mean().idxmax()
```

[25]: 'Bollinger Pink platted moscato rose'

5: What is the top-selling product category in terms of quantity sold or revenue generated?

```
[26]: df.groupby('Sub_Category').Amount.mean().idxmax()
```

[26]: 'Champagne & Spakling Wine'

Conclusion (4,5): One can see that Bollinger Pink platted moscato rose is the product that generates the highest revenue. Also, one can see that Champagne & Sparkling Wine is the top-selling producct category.

Business ideas: 1. Wine Subscription Service: Launch a subscription service specializing in premium rosé wines, including Bollinger Pink platted moscato Rosé.

2. Wine Tasting Tours: If you're located in a wine-producing region, consider offering wine tasting tours that focus on rosé wines, including a tasting of Bollinger Pink Platted Moscato Rosé. Create unique and immersive experiences by partnering with local wineries for tours and tastings.

6: Who buy more in the "Frozen" category: men or women?

```
[29]: gender_counts = df[df['Category'] == 'Frozen']['Gender'].value_counts()
print(gender_counts)
```

Gender

Female 5401 Male 992

Name: count, dtype: int64

Conclusion: There less of male customers in the "Frozen" category.

Business idea: Product Placement and Store Layout: Adjust the store layout to prominently display frozen food items that are popular among men. Placing these products in areas that men frequently visit can increase their visibility and encourage more purchases.

7: At which category do we have the most expensive purchase?

```
[47]: df.groupby('Membership ID').Price.mean().idxmax()
df.loc[df['Membership ID'] == 'MIDSG0094'].Category
```

[47]: 32676 Mother & Baby
Name: Category, dtype: object

Conclusion: We have the most expensive purchase in the category: "Mother & Baby".

Business idea: Baby Shower Planning Services: Offer event planning services for baby showers, gender reveal parties, and other related celebrations. Create memorable experiences for expectant parents and their friends and family.

8: At what city customers have the highest average bill?

```
[76]: df.groupby('City').Amount.mean()
```

[76]: City

Bedok 11.054826

Jakarta 11.310233

Kuala Lumpur 11.262599

Manila 10.822900

Woodlands 11.732765

Name: Amount, dtype: float64

Conclusion: In Woodlands customers have the highest average bill.

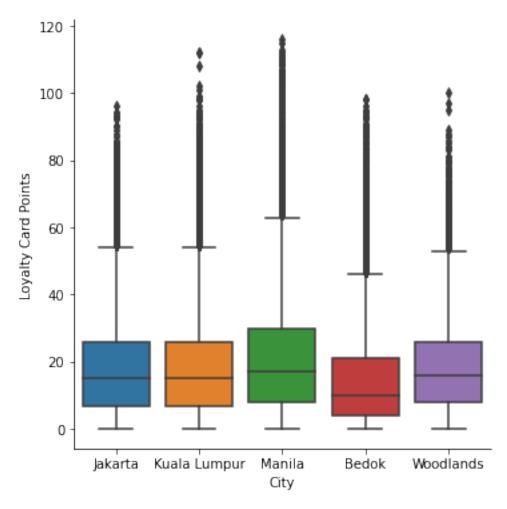
Business idea: Exclusive Cooking Classes: Organize exclusive cooking classes and workshops that teach customers how to prepare gourmet meals using premium ingredients. Collaborate with renowned chefs or culinary experts to provide a unique and educational experience. Offer classes in-person and online to reach a wider audience.

10: At what city people have the lowest level of loyalty card points?

```
[19]: plot = sns.catplot(data = df, x="City", y="Loyalty Card Points", kind="box") plot.fig.suptitle("Loyalty Card Points by Cities", y=1.05) plot
```

[19]: <seaborn.axisgrid.FacetGrid at 0x7fbf63af4580>





Conclusion: In Bedock people have the lowest level of loyalty card points.

Business idea: Loyalty Point Boost Program: Create a loyalty point boosting program specifically for Bedeck grocery stores.

Let us create a pivot table

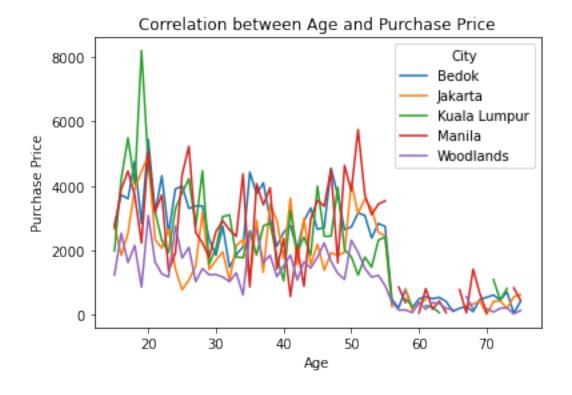
```
[9]: City
               Bedok
                        Jakarta
                                 Kuala Lumpur
                                                  Manila Woodlands
    Age
           2784.5174
     15
                      2793.6930
                                    1979.3798
                                               2684.6992
                                                          1231.9019
                                                          2529.8478
     16
           3723.5775
                     1839.1363
                                    4214.1807
                                               3890.7314
     17
           3601.8660 2530.0170
                                    5495.0694 4463.0577
                                                           1619.7760
     18
           4770.9461
                     3918.6245
                                    4049.1765
                                               3696.5842
                                                          2151.6184
     19
           2830.9589 4447.4399
                                    8208.4761 2224.9030
                                                           860.2747
```

11: What is the corelation between age and purchase price for every city?

```
[16]: ap = age_pivot.plot()
    ap.set_xlabel("Age")
    ap.set_ylabel("Purchase Price")

ap.set_title("Correlation between Age and Purchase Price")
```

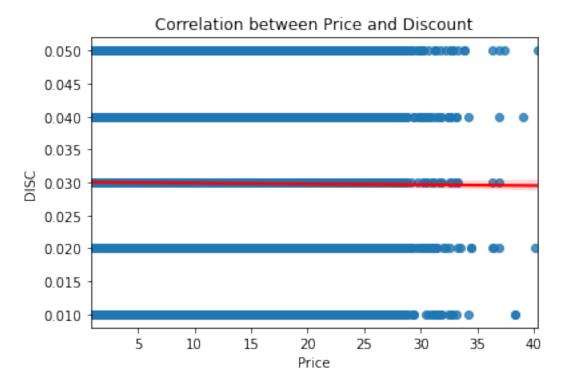
[16]: Text(0.5, 1.0, 'Correlation between Age and Purchase Price')



Conclusion: one sees that in Kuala Lumpur people around 20 make the most expensive purchases. That look suspicious and takes needs some additional reserch to be done.

13: Is there any correlation between Price and Discount?

```
[25]: sns.regplot(
    data = df,
    x = 'Price', y='DISC', line_kws = {'color': 'red'}
);
plt.title("Correlation between Price and Discount")
plt.show()
```



Conclusion: There is a direct correlation between Price and Discount. No outliers.

Let us create another pivot table

```
[34]: gender_pivot = df.pivot_table(index='Gender', columns='City', u

→values='Amount', aggfunc = 'sum')
gender_pivot.head()
```

```
[34]: City
                               Jakarta Kuala Lumpur
                    Bedok
                                                            Manila
                                                                     Woodlands
      Gender
     Female
              122601.1731
                            83361.1450
                                          92527.7172
                                                       104309.6095
                                                                    66404.0654
     Male
               13063.6490
                            19652.4567
                                          28691.6373
                                                        29677.8889
                                                                     3100.8349
```

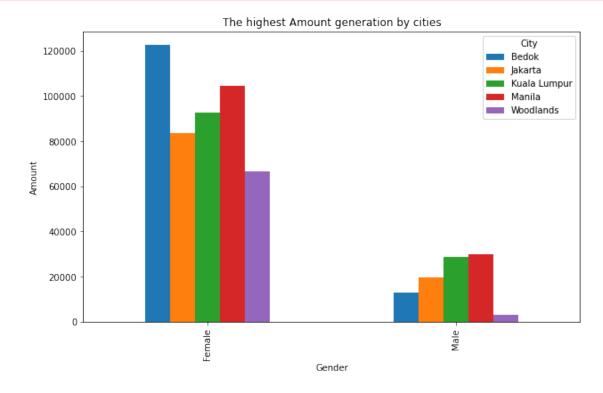
14: At what city do female customers generate the highest Amount?

```
[35]: ax = gender_pivot.plot.bar(figsize=(10,6));
ax.set_title("The highest Amount generation by cities")
ax.set_ylabel("Amount")
ax.show()
```

```
AttributeError Traceback (most recent call last)

<ipython-input-35-1a481d89051c> in <module>
        2 ax.set_title("The highest Amount generation by cities")
        3 ax.set_ylabel("Amount")
----> 4 ax.show()

AttributeError: 'AxesSubplot' object has no attribute 'show'
```



Conclusion: in Bedok female customers generate the highest Amount

Business ideas: 1.Inventory and Stocking: If female customers in Bedok prefer specific products, the business can optimize its inventory and stocking decisions. It can ensure that the products favored by this demographic are well-stocked and readily available.

2. Feedback and Surveys: To gain deeper insights into why female customers in Bedok are spending more, the business can conduct surveys or gather feedback. This information can help refine strategies and offerings.

15: What are the average pre-tax checks for each subcategory in the beauty and care

category? Let's create a pie chart and review the structure

```
[7]: beaty_net_bill = df[df['Category'] == 'Beauty'].groupby('Sub_Category')['Net_

→Bill Amount'].sum()

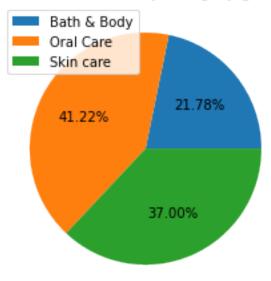
beaty_net_bill
```

[7]: Sub_Category
Bath & Body 84351.7905
Oral Care 159649.8545
skin care 143328.2248

Name: Net Bill Amount, dtype: float64

```
[9]: plt.pie(beaty_net_bill, autopct = '%1.2f%%')
plt.title('Structure of Beauty category goods')
plt.legend(labels = ['Bath & Body', 'Oral Care', 'Skin care']);
```

Structure of Beauty category goods



Conclusion: The oral care goods have the highest pre-tax checks.