

# sales-marketing-campaign

November 5, 2024

```
[1]: import pandas as pd
data = pd.read_csv('C:\\Users\\ANUSHA\\Downloads\\sales_marketing_campaign_data.
↳csv')
```

```
[22]: print(data.head())
print(data.info())
print(data.describe())
```

	Customer_ID	Campaign_ID	Region	Age	Gender	Income_Level	\
0	1	Campaign_A	North	60	Male	Low	
1	2	Campaign_B	West	38	Male	Medium	
2	3	Campaign_A	West	23	Female	Medium	
3	4	Campaign_B	East	57	Female	Low	
4	5	Campaign_A	South	24	Male	Medium	

	Sales_Before_Campaign	Sales_After_Campaign	Engagement_Level	\
0	913.39	778.45	8	
1	565.49	901.61	9	
2	125.65	1192.04	6	
3	686.29	766.30	3	
4	808.22	126.44	4	

	Purchase_Frequency	Age_Group
0	4	46-65
1	2	31-45
2	3	18-30
3	4	46-65
4	1	18-30

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 100 entries, 0 to 99

Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	Customer_ID	100 non-null	int64
1	Campaign_ID	100 non-null	object
2	Region	100 non-null	object
3	Age	100 non-null	int64
4	Gender	100 non-null	object

```

5   Income_Level          100 non-null   object
6   Sales_Before_Campaign 100 non-null   float64
7   Sales_After_Campaign  100 non-null   float64
8   Engagement_Level      100 non-null   int64
9   Purchase_Frequency    100 non-null   int64
10  Age_Group             100 non-null   category
dtypes: category(1), float64(2), int64(4), object(4)
memory usage: 8.2+ KB
None

```

	Customer_ID	Age	Sales_Before_Campaign	Sales_After_Campaign \
count	100.000000	100.000000	100.00000	100.00000
mean	50.500000	40.030000	549.01510	872.16110
std	29.011492	13.411479	252.02248	398.42552
min	1.000000	19.000000	101.15000	126.44000
25%	25.750000	28.750000	347.31500	554.58750
50%	50.500000	41.500000	540.67500	894.50500
75%	75.250000	50.250000	725.65500	1256.27500
max	100.000000	64.000000	998.92000	1493.57000

	Engagement_Level	Purchase_Frequency
count	100.000000	100.000000
mean	5.100000	3.080000
std	2.830212	1.375691
min	1.000000	1.000000
25%	2.750000	2.000000
50%	5.000000	3.000000
75%	7.000000	4.000000
max	10.000000	5.000000

```
[3]: print(data.isnull().sum())
```

```

Customer_ID          0
Campaign_ID          0
Region              0
Age                 0
Gender              0
Income_Level        0
Sales_Before_Campaign 0
Sales_After_Campaign 0
Engagement_Level    0
Purchase_Frequency  0
dtype: int64

```

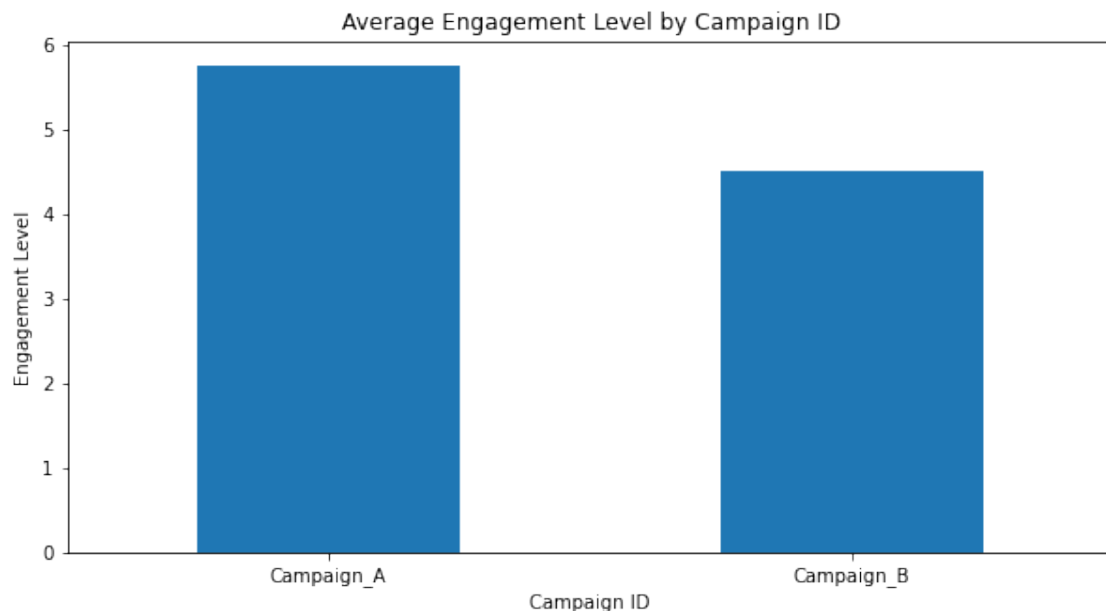
```

[20]: # Engagement level by Campaign ID
engagement_comparison = data.groupby('Campaign_ID')['Engagement_Level'].mean()
print(engagement_comparison)
engagement_comparison.plot(kind='bar', figsize=(10, 5))

```

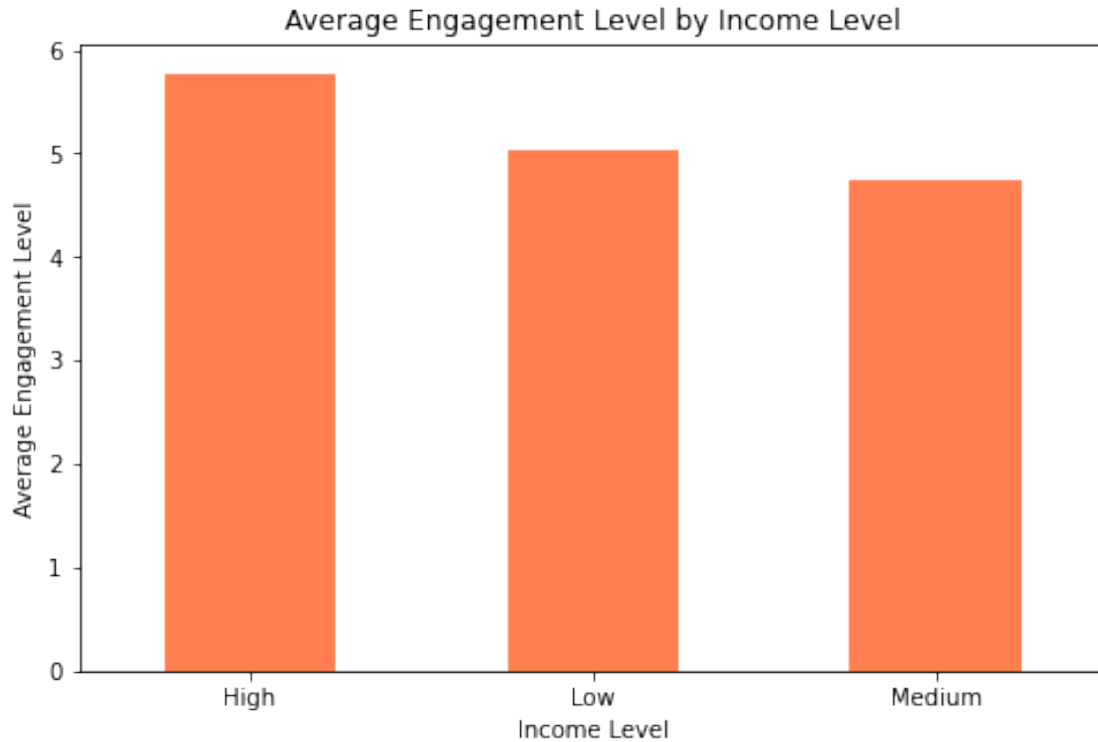
```
plt.title('Average Engagement Level by Campaign ID')
plt.ylabel('Engagement Level')
plt.xlabel('Campaign ID')
plt.xticks(rotation=0)
plt.show()
```

```
Campaign_ID
Campaign_A    5.75
Campaign_B    4.50
Name: Engagement_Level, dtype: float64
```



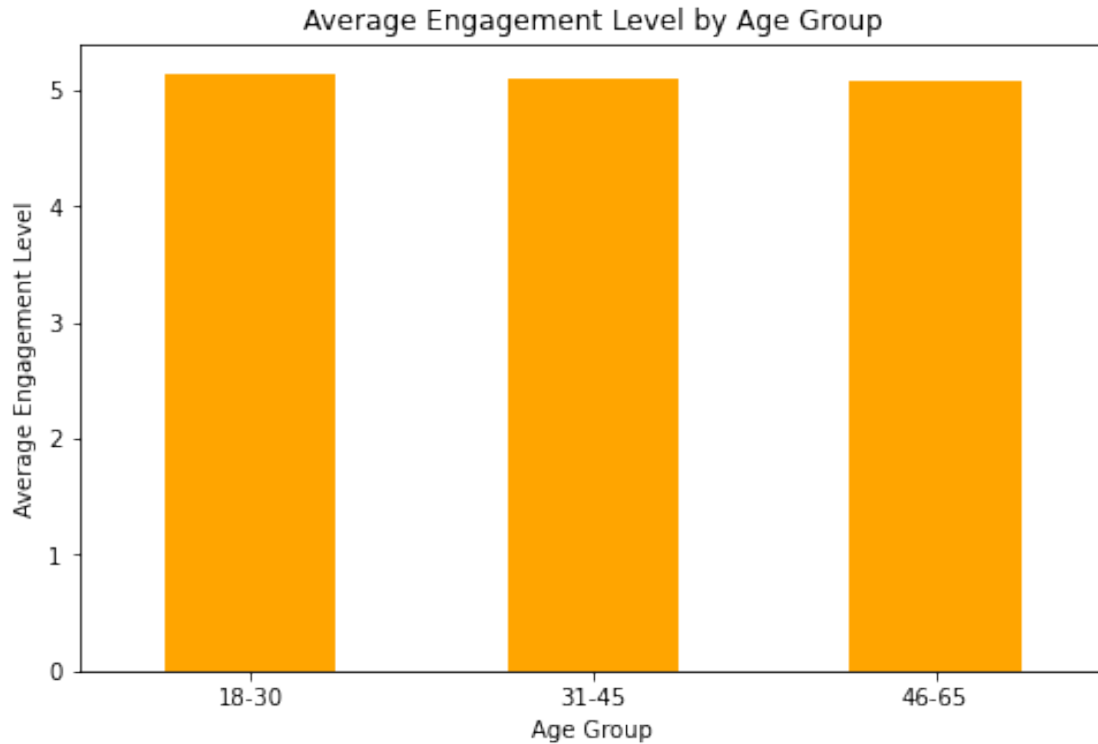
```
[14]: # Engagement level by income level
engagement_by_income = data.groupby('Income_Level')['Engagement_Level'].mean()
print(engagement_by_income)
engagement_by_income.plot(kind='bar', color='coral', figsize=(8, 5))
plt.title('Average Engagement Level by Income Level')
plt.xlabel('Income Level')
plt.ylabel('Average Engagement Level')
plt.xticks(rotation=0)
plt.show()
```

```
Income_Level
High    5.769231
Low     5.032258
Medium  4.744186
Name: Engagement_Level, dtype: float64
```



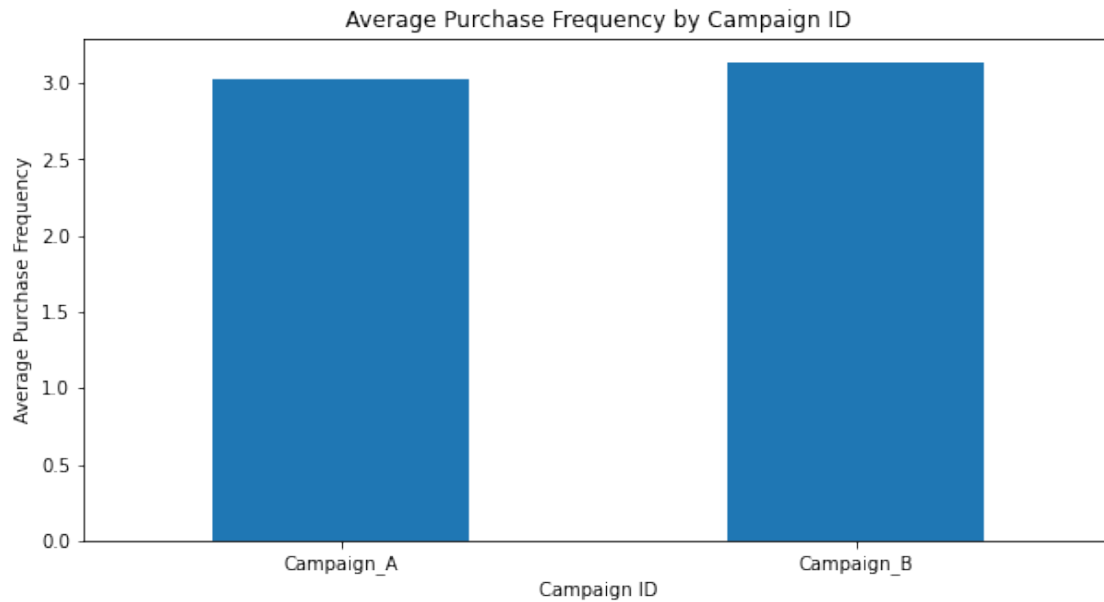
```
[23]: bins = [18, 30, 45, 65]
labels = ['18-30', '31-45', '46-65']
data['Age_Group'] = pd.cut(data['Age'], bins=bins, labels=labels, right=False)
# Calculate mean engagement level by age group
engagement_by_age_group = data.groupby('Age_Group')['Engagement_Level'].mean()
print(engagement_by_age_group)
engagement_by_age_group.plot(kind='bar', color='orange', figsize=(8, 5))
plt.title('Average Engagement Level by Age Group')
plt.xlabel('Age Group')
plt.ylabel('Average Engagement Level')
plt.xticks(rotation=0)
plt.show()
```

```
Age_Group
18-30    5.142857
31-45    5.096774
46-65    5.073171
Name: Engagement_Level, dtype: float64
```



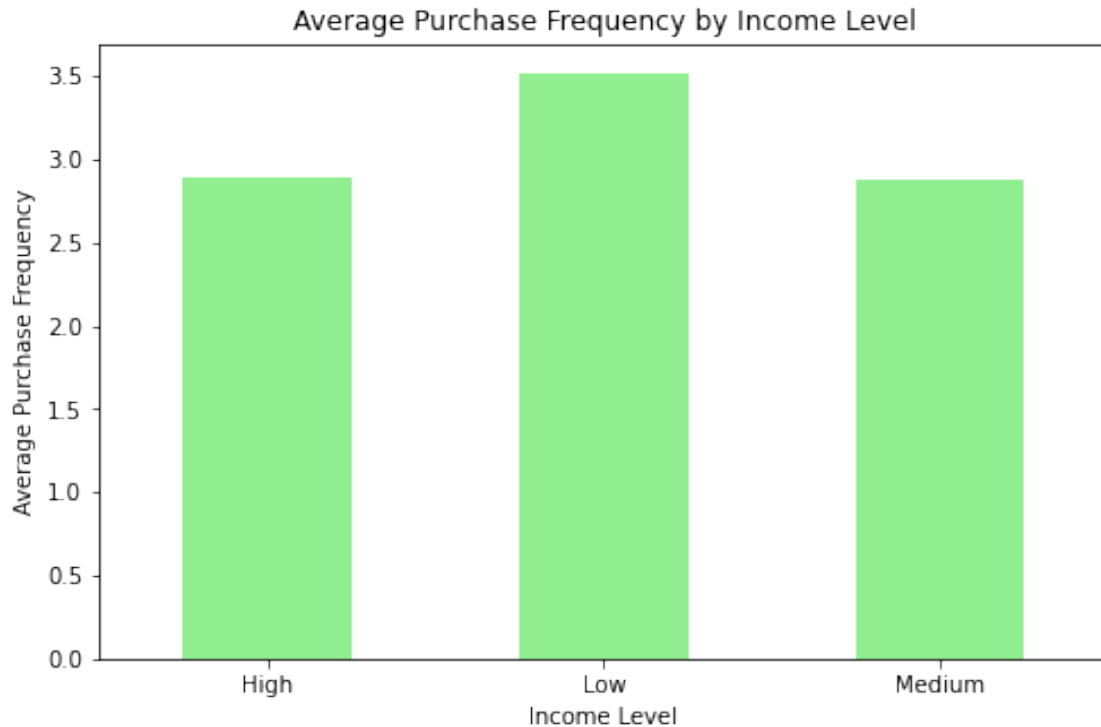
```
[17]: # purchase frequency by Campaign ID
purchase_frequency_comparison = data.
      ↳groupby('Campaign_ID')['Purchase_Frequency'].mean()
print(purchase_frequency_comparison)
purchase_frequency_comparison.plot(kind='bar', figsize=(10, 5))
plt.title('Average Purchase Frequency by Campaign ID')
plt.ylabel('Average Purchase Frequency')
plt.xlabel('Campaign ID')
plt.xticks(rotation=0)
plt.show()
```

```
Campaign_ID
Campaign_A    3.020833
Campaign_B    3.134615
Name: Purchase_Frequency, dtype: float64
```



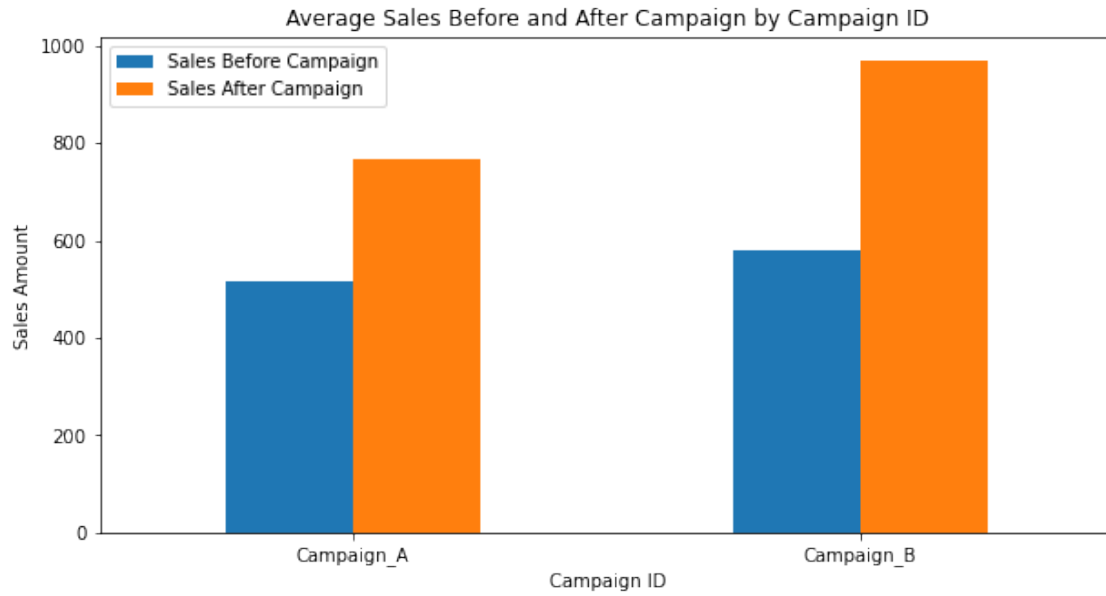
```
[13]: # purchase frequency by income level
purchase_freq_by_income = data.groupby('Income_Level')['Purchase_Frequency'].
    ↪mean()
print(purchase_freq_by_income)
purchase_freq_by_income.plot(kind='bar', color='lightgreen', figsize=(8, 5))
plt.title('Average Purchase Frequency by Income Level')
plt.xlabel('Income Level')
plt.ylabel('Average Purchase Frequency')
plt.xticks(rotation=0)
plt.show()
```

```
Income_Level
High      2.884615
Low       3.516129
Medium    2.883721
Name: Purchase_Frequency, dtype: float64
```



```
[25]: sales_comparison = data.groupby('Campaign_ID')[['Sales_Before_Campaign', 'Sales_After_Campaign']].mean()
print(sales_comparison)
import matplotlib.pyplot as plt
sales_comparison.plot(kind='bar', figsize=(10, 5))
plt.title('Average Sales Before and After Campaign by Campaign ID')
plt.ylabel('Sales Amount')
plt.xlabel('Campaign ID')
plt.xticks(rotation=0)
plt.legend(['Sales Before Campaign', 'Sales After Campaign'])
plt.show()
```

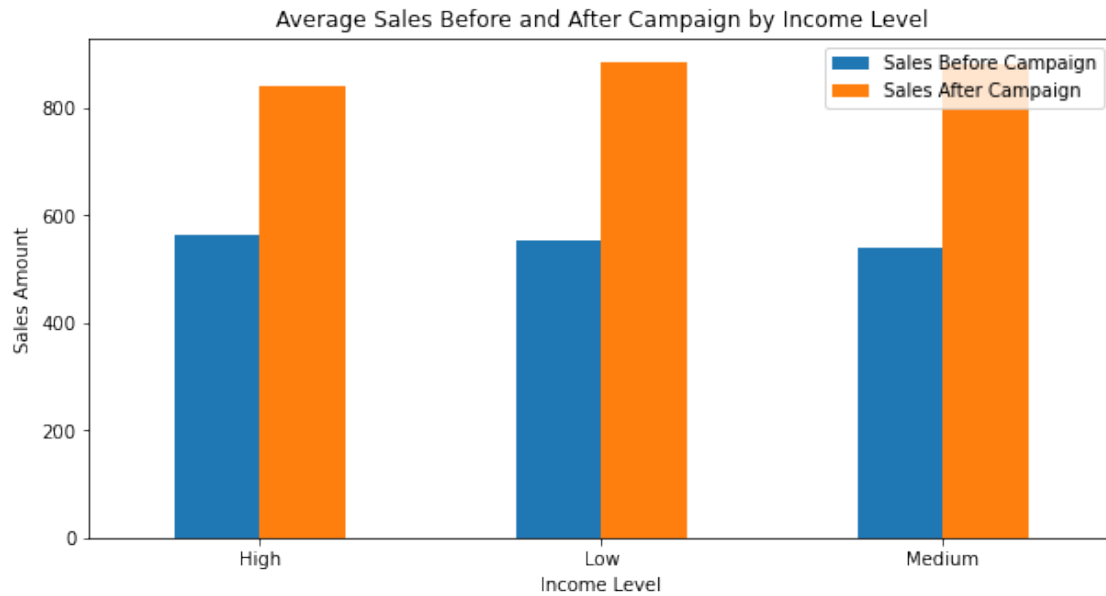
	Sales_Before_Campaign	Sales_After_Campaign
Campaign_ID		
Campaign_A	516.548750	767.348333
Campaign_B	578.984038	968.911346



```
[18]: income_sales_comparison = data.
      ↳groupby('Income_Level')[['Sales_Before_Campaign', 'Sales_After_Campaign']].
      ↳mean()
      print(income_sales_comparison)
      income_sales_comparison.plot(kind='bar', figsize=(10, 5))
      plt.title('Average Sales Before and After Campaign by Income Level')
      plt.ylabel('Sales Amount')
      plt.xlabel('Income Level')
      plt.xticks(rotation=0)
      plt.legend(['Sales Before Campaign', 'Sales After Campaign'])
      plt.show()
```

Income_Level	Sales_Before_Campaign	Sales_After_Campaign
High	562.165000	842.155000
Low	552.051613	885.094839
Medium	538.874884	880.980000





```
[15]: from scipy.stats import f_oneway

# Extract engagement levels for each income group
high_engagement = data[data['Income_Level'] == 'High']['Engagement_Level']
medium_engagement = data[data['Income_Level'] == 'Medium']['Engagement_Level']
low_engagement = data[data['Income_Level'] == 'Low']['Engagement_Level']

# ANOVA test
f_stat, p_val = f_oneway(high_engagement, medium_engagement, low_engagement)
print("ANOVA test for Engagement Level by Income Level: F-statistic =", f_stat,
      "\u2192", p-value =", p_val)

#p-value < 0.05, it suggests significant differences in engagement levels
\u2192 across income groups.
```

ANOVA test for Engagement Level by Income Level: F-statistic =  
1.0772471187142683 , p-value = 0.3445700572160334