

## AI-Enabled Marketing Campaign Analysis using Pandas

```
import pandas as pd
```

### Introduction - Dataset Summary

```
df= pd.read_csv('/content/drive/MyDrive/Datasets/marketing_campaign_sample.csv') #reading dataset contains marketing campaign data
```

```
#Analyzing its structure and summary stats
```

```
print(df.head()) #head reads first 5 rows of dataset
```

```
↵
```

|   | Campaign   | Impressions | Clicks | Channel | Spend |
|---|------------|-------------|--------|---------|-------|
| 0 | Campaign A | 10000       | 500    | Email   | 200   |
| 1 | Campaign B | 25000       | 1200   | Social  | 450   |
| 2 | Campaign C | 15000       | 750    | Search  | 300   |
| 3 | Campaign D | 12000       | 600    | Email   | 250   |
| 4 | Campaign E | 30000       | 1000   | Social  | 500   |

```
#campaign : Name of the marketing campaign
#impressions : Number of times the ad was shown
#clicks : Number of times the users clicked the ad
#channel : Platform where campaign ran (Email,social,search)
#spend : Total amount spent on campaign
```

```
df.columns #To find column names of dataframes
```

```
↵ Index(['Campaign', 'Impressions', 'Clicks', 'Channel', 'Spend'], dtype='object')
```

```
df.shape #prints number of rows and cloumns
```

```
↵ (15, 5)
```

```
df.info() #To find Data type,nullvalues
```

```
↵ <class 'pandas.core.frame.DataFrame'>
RangeIndex: 15 entries, 0 to 14
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Campaign    15 non-null    object
1   Impressions 15 non-null    int64
2   Clicks      15 non-null    int64
3   Channel     15 non-null    object
4   Spend       15 non-null    int64
dtypes: int64(3), object(2)
memory usage: 732.0+ bytes
```

```
#df.describe() #summmary
```

```
df['Channel'].unique()
```

```
↵ array(['Email', 'Social', 'Search'], dtype=object)
```

```
df['Channel'].value_counts() #To know which channel used most for marketing campaigns
```

```
↵
```

| Channel | count |
|---------|-------|
| Social  | 6     |
| Search  | 5     |
| Email   | 4     |

```
df.dtypes
```

*#DAY 1 - Analyzed the data to understand what it represents, examined each column to identify its meaning, checked column names, detected any null values, verified data types, and calculated basic statistics such as mean and median*

```
#DAY2
```

```
df.isnull().sum() #it checks how many null values are present in each column
```

```
0
Campaign    0
Impressions 0
Clicks      0
Channel     0
Spend       0
```

```
df.duplicated().sum() #checks there is any duplicate values. 0 in the case of no duplicates
```

```
np.int64(0)
```

```
df.dtypes #checks the datatypes of each column
```

```
0
Campaign    object
Impressions int64
Clicks      int64
Channel     object
Spend       int64
```

```
df['CTR']=df['Clicks']/df['Impressions'] *100 #It adds new columns to the df
```

```
df.rename(columns={'Spend':'Spend (INR)'},inplace=True)
```

```
df
```


```
0
Campaign Impressions Clicks Channel Spend (INR) CTR
0 Campaign A      10000    500   Email         200  5.000000
1 Campaign B      25000   1200   Social         450  4.800000
2 Campaign C      15000    750   Search         300  5.000000
3 Campaign D      12000    600   Email         250  5.000000
4 Campaign E      30000   1000   Social         500  3.333333
5 Campaign F       5000    200   Search         100  4.000000
6 Campaign G      10191   362    Social         256  3.552154
7 Campaign H       5952   184    Search         291  3.091398
8 Campaign I      13064   269    Email         400  2.059094
9 Campaign J      13034  1099    Social         332  8.431794
10 Campaign K     14904   687    Search         180  4.609501
11 Campaign L     13885  1073    Social         251  7.727764
12 Campaign M     10374   662    Search         246  6.381338
13 Campaign N       5050   554    Email         196 10.970297
14 Campaign O     12816   652    Social         371  5.087391
```

Next steps:



[Generate code with df](#)
[View recommended plots](#)
[New interactive sheet](#)

```
#No missing values, 0 duplicates, CTR column added
```


```
df.sort_values('CTR',ascending=False)
```



|    | Campaign   | Impressions | Clicks | Channel | Spend (INR) | CTR       |
|----|------------|-------------|--------|---------|-------------|-----------|
| 13 | Campaign N | 5050        | 554    | Email   | 196         | 10.970297 |
| 9  | Campaign J | 13034       | 1099   | Social  | 332         | 8.431794  |
| 11 | Campaign L | 13885       | 1073   | Social  | 251         | 7.727764  |
| 12 | Campaign M | 10374       | 662    | Search  | 246         | 6.381338  |
| 14 | Campaign O | 12816       | 652    | Social  | 371         | 5.087391  |
| 2  | Campaign C | 15000       | 750    | Search  | 300         | 5.000000  |
| 0  | Campaign A | 10000       | 500    | Email   | 200         | 5.000000  |
| 3  | Campaign D | 12000       | 600    | Email   | 250         | 5.000000  |
| 1  | Campaign B | 25000       | 1200   | Social  | 450         | 4.800000  |
| 10 | Campaign K | 14904       | 687    | Search  | 180         | 4.609501  |
| 5  | Campaign F | 5000        | 200    | Search  | 100         | 4.000000  |
| 6  | Campaign G | 10191       | 362    | Social  | 256         | 3.552154  |
| 4  | Campaign E | 30000       | 1000   | Social  | 500         | 3.333333  |
| 7  | Campaign H | 5952        | 184    | Search  | 291         | 3.091398  |
| 8  | Campaign I | 13064       | 269    | Email   | 400         | 2.059094  |


```
df['CTR'].describe() #for insight
```




|       | CTR       |
|-------|-----------|
| count | 15.000000 |
| mean  | 5.269604  |
| std   | 2.299783  |
| min   | 2.059094  |
| 25%   | 3.776077  |
| 50%   | 5.000000  |
| 75%   | 5.734364  |
| max   | 10.970297 |

```
#DAY3
```


```
df.groupby('Channel')[['Impressions','Clicks','Spend (INR)','CTR']].mean() #We grouped channels based on impressions,clicks,Spend(INR)
```



|         | Impressions  | Clicks     | Spend (INR) | CTR      |
|---------|--------------|------------|-------------|----------|
| Channel |              |            |             |          |
| Email   | 10028.500000 | 480.750000 | 261.5       | 5.757348 |
| Search  | 10246.000000 | 496.600000 | 223.4       | 4.616447 |
| Social  | 17487.666667 | 897.666667 | 360.0       | 5.488739 |



```
df.groupby('Channel')['CTR'].mean().sort_values(ascending=False) #We grouped channels based on CTR
```



|         | CTR      |
|---------|----------|
| Channel |          |
| Email   | 5.757348 |
| Social  | 5.488739 |
| Search  | 4.616447 |

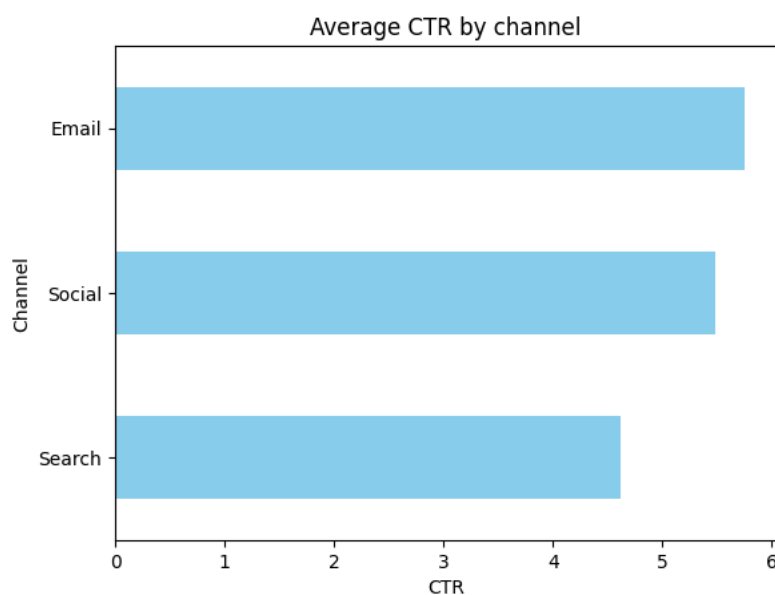
```
df.groupby('Campaign')[['Impressions','Clicks','Spend (INR)','CTR']].sum() #Grouping campaign based on impressions,clicks,spend
```



|            | Impressions | Clicks | Spend (INR) | CTR       |
|------------|-------------|--------|-------------|-----------|
| Campaign   |             |        |             |           |
| Campaign A | 10000       | 500    | 200         | 5.000000  |
| Campaign B | 25000       | 1200   | 450         | 4.800000  |
| Campaign C | 15000       | 750    | 300         | 5.000000  |
| Campaign D | 12000       | 600    | 250         | 5.000000  |
| Campaign E | 30000       | 1000   | 500         | 3.333333  |
| Campaign F | 5000        | 200    | 100         | 4.000000  |
| Campaign G | 10191       | 362    | 256         | 3.552154  |
| Campaign H | 5952        | 184    | 291         | 3.091398  |
| Campaign I | 13064       | 269    | 400         | 2.059094  |
| Campaign J | 13034       | 1099   | 332         | 8.431794  |
| Campaign K | 14904       | 687    | 180         | 4.609501  |
| Campaign L | 13885       | 1073   | 251         | 7.727764  |
| Campaign M | 10374       | 662    | 246         | 6.381338  |
| Campaign N | 5050        | 554    | 196         | 10.970297 |
| Campaign O | 12816       | 652    | 371         | 5.087391  |

#VISUALISATION

```
import matplotlib.pyplot as plt
df.groupby('Channel')['CTR'].mean().sort_values().plot(kind='barh',color='skyblue',)
plt.title('Average CTR by channel')
plt.xlabel('CTR')
plt.ylabel('Channel')
plt.show()
```



#Email has highest CTR  
 #Based on money spend, Campaign E is on Top

#DAY 4


#Visualize and Communicate like a pro !

#1. BAR CHART AND CTR% By Campaign

```
import matplotlib.pyplot as plt
import seaborn as sns
```

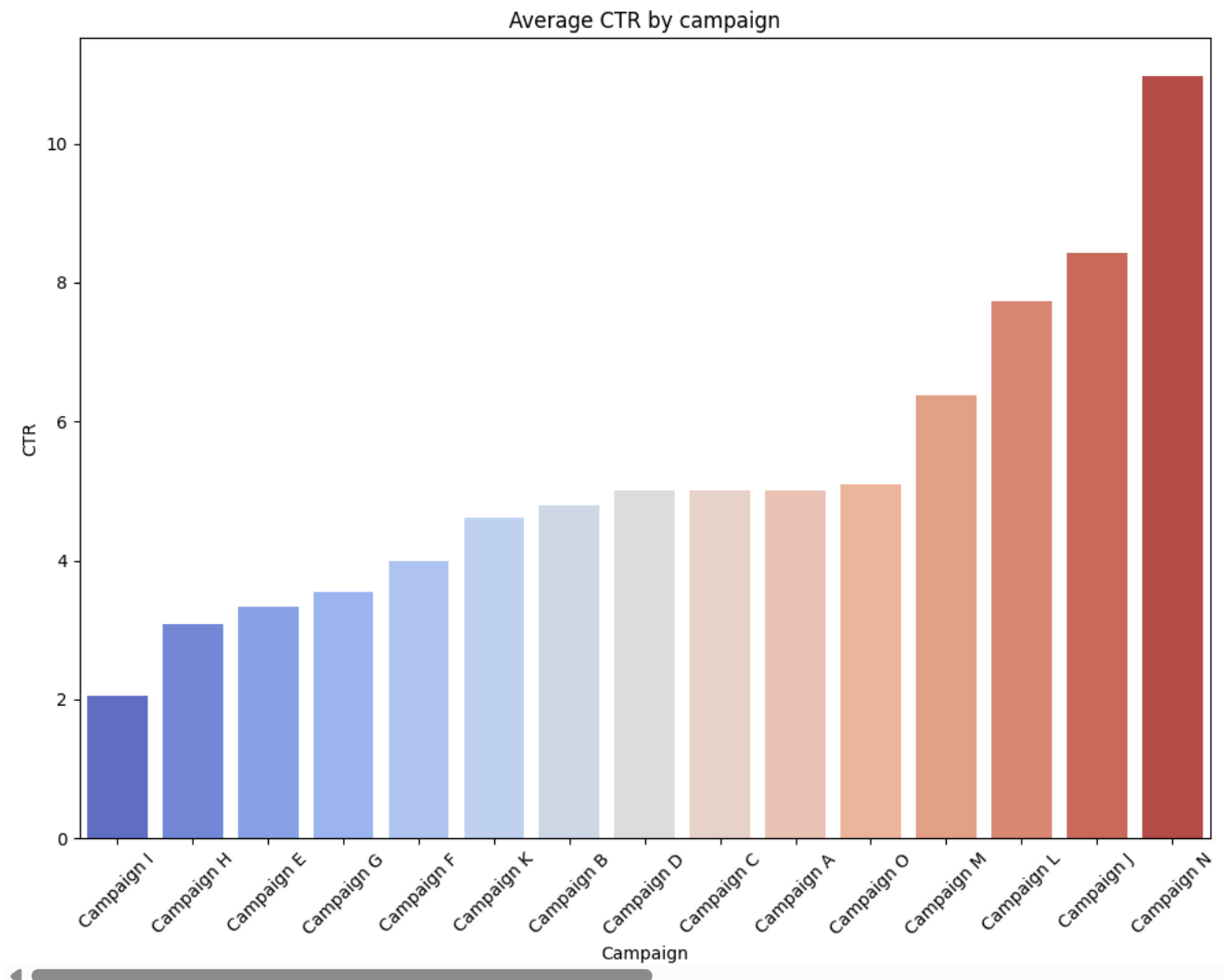
```
campaign_ctr = df.groupby('Campaign')['CTR'].mean().sort_values()
```

```
#plot bar chart
plt.figure(figsize=(10,8))
sns.barplot(x=campaign_ctr.index,y=campaign_ctr.values,palette='coolwarm')
plt.title("Average CTR by campaign")
plt.xlabel("Campaign")
plt.ylabel("CTR")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

 <ipython-input-69-1333971358>:8: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le

```
sns.barplot(x=campaign_ctr.index,y=campaign_ctr.values,palette='coolwarm')
```



#Campaign 'N' delivered the highest CTR,suggesting more effective targeting or messaging.

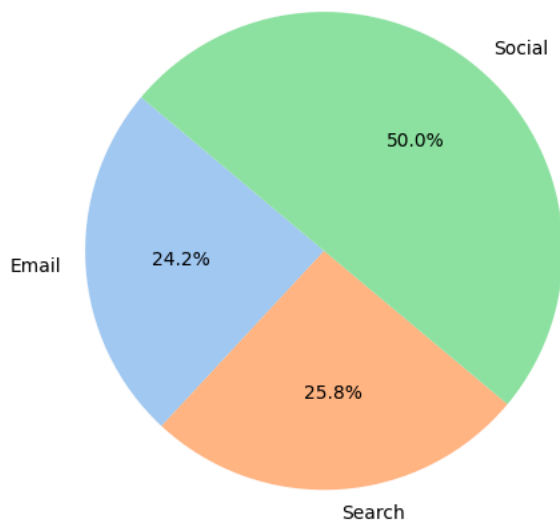
Total Spend By channel

```
channel_spend = df.groupby('Channel')['Spend (INR)'].sum().sort_values()

plt.figure(figsize=(8,5))
plt.pie(channel_spend,labels=channel_spend.index,autopct='%1.1f%%',startangle=140,colors=sns.color_palette('pastel'))
plt.title('Spend distribution by channel')
plt.tight_layout()
plt.show()
```



Spend distribution by channel



#"Social received the largest share of ad spend."

Start coding or [generate](#) with AI.

\*"Social is leading in Impressions, Clicks, and Spend, indicating that investing more in Social was a good decision. However, Email has the highest CTR, suggesting that focusing more on Email could yield excellent results due to its high ad click-through rate."\*

#### #FINAL INSIGHTS

#Camapaign N had the highest CTR (10.97%)

#Email channel had the best engagement (CTR 5.75%)

#Social had highest spend,but Email delivered better results

#-Tools used

#- Python, Pandas,Matplotlib,Seaborn