

```
In [1]: import pandas as pd
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
In [2]: df = pd.read_csv(r"C:\Users\santo\OneDrive\Desktop\Data science\Dec2025\14th dec EDA\social_media_engagement1.csv")
```

```
In [3]: df.head()
```

	post_id	platform	post_type	post_time	likes	comments	shares	post_day	sentiment_score
0	1	Facebook	image	8/17/2023 14:45	2121	474	628	Thursday	positive
1	2	Facebook	carousel	5/14/2023 0:45	3660	432	694	Sunday	neutral
2	3	Instagram	poll	2/21/2023 16:15	4955	408	688	Tuesday	negative
3	4	Twitter	image	11/16/2023 0:45	1183	90	187	Thursday	negative
4	5	Twitter	video	5/23/2023 0:30	3499	247	286	Tuesday	positive

```
In [6]: ("\\nData Info:\\n", df.info)
```

```
Out[6]: ('\\nData Info:\\n',
      <bound method DataFrame.info of      post_id  platform post_type      post_time  likes  comments  shares  \\
      0        1  Facebook    image  8/17/2023 14:45    2121      474     628
      1        2  Facebook  carousel  5/14/2023 0:45    3660      432     694
      2        3  Instagram   poll  2/21/2023 16:15    4955      408     688
      3        4  Twitter    image  11/16/2023 0:45   1183       90     187
      4        5  Twitter    video  5/23/2023 0:30   3499      247     286
      ..      ...
      95       96  Instagram  carousel  7/12/2023 17:45      36     294     911
      96       97  Twitter    video  10/27/2023 23:45    314      108     458
      97       98  Twitter    text   8/5/2023 8:45    229      179      38
      98       99  Instagram   poll  12/29/2023 12:15   5000      500     204
      99      100  Instagram   image  6/6/2023 21:00   4483      357      25

      post_day sentiment_score
      0  Thursday      positive
      1  Sunday        neutral
      2  Tuesday       negative
      3  Thursday       negative
      4  Tuesday       positive
      ..      ...
      95 Wednesday      positive
      96  Friday        neutral
      97 Saturday       positive
      98  Friday        positive
      99  Tuesday       neutral

      [100 rows x 9 columns]>)
```

```
In [7]: df.describe()
```

Out[7]:

	post_id	likes	comments	shares
<b>count</b>	100.000000	100.000000	100.000000	100.000000
<b>mean</b>	50.500000	2381.810000	202.660000	415.650000
<b>std</b>	29.011492	1632.573284	138.84067	283.877601
<b>min</b>	1.000000	15.000000	10.00000	16.000000
<b>25%</b>	25.750000	895.750000	89.75000	183.000000
<b>50%</b>	50.500000	2220.000000	171.00000	356.500000
<b>75%</b>	75.250000	3593.250000	299.00000	689.500000
<b>max</b>	100.000000	5000.000000	500.00000	993.000000

## Data Scrubbing

In [9]: `print(df.isnull().sum())`

```
post_id      0
platform     0
post_type    0
post_time    0
likes        0
comments     0
shares        0
post_day     0
sentiment_score  0
dtype: int64
```

In [10]: `# Check missing values
print(df.isnull().sum())`

```
# Option 1: Drop rows with missing values
df = df.dropna()
```

```
# Option 2: Fill missing values (example: fill numeric with 0, categorical with mode)
df["likes"] = df["likes"].fillna(0)
df["comments"] = df["comments"].fillna(0)
df["shares"] = df["shares"].fillna(0)

for col in ["platform", "post_type", "post_day", "sentiment_score"]:
    df[col] = df[col].fillna(df[col].mode()[0])
```

```
post_id      0
platform     0
post_type    0
post_time    0
likes        0
comments     0
shares        0
post_day     0
sentiment_score 0
dtype: int64
```

checking Duplicate Values:

```
In [11]: print("Duplicate rows:", df.duplicated().sum())
df = df.drop_duplicates()
```

```
Duplicate rows: 0
```

## Data Exploration

Summary of the data

```
In [13]: summary_platform = df.groupby("platform")[["likes", "comments", "shares"]].mean().round(3) #mean is rounded with three decimal

summary_platform
```

Out[13]:

**likes comments shares**

platform	likes	comments	shares
Facebook	2699.750	248.906	474.375
Instagram	2999.833	232.444	525.167
Twitter	1368.594	122.906	233.719

In [14]:

#Summary of engagement by post type:

```
summary_post_type = df.groupby("post_type")[["likes", "comments", "shares"]].mean().round(3) #mean is rounded with three decimal points
```

summary\_post\_type

Out[14]:

**likes comments shares**

post_type	likes	comments	shares
carousel	2263.577	236.000	531.154
image	2032.765	197.176	371.529
poll	3061.077	214.538	470.769
text	1815.286	138.381	274.952
video	2906.783	221.000	415.000

In [15]:

# Create engagement column

```
df["engagement"] = df["likes"] + df["comments"] + df["shares"]
```

# Average engagement per platform

```
engagement_platform = df.groupby("platform")["engagement"].mean().round(2) #mean is rounded with two decimal points
```

engagement\_platform

```
Out[15]: platform
          Facebook    3423.03
          Instagram   3757.44
          Twitter     1725.22
          Name: engagement, dtype: float64
```

## iNterpretation (Visualization of the data)

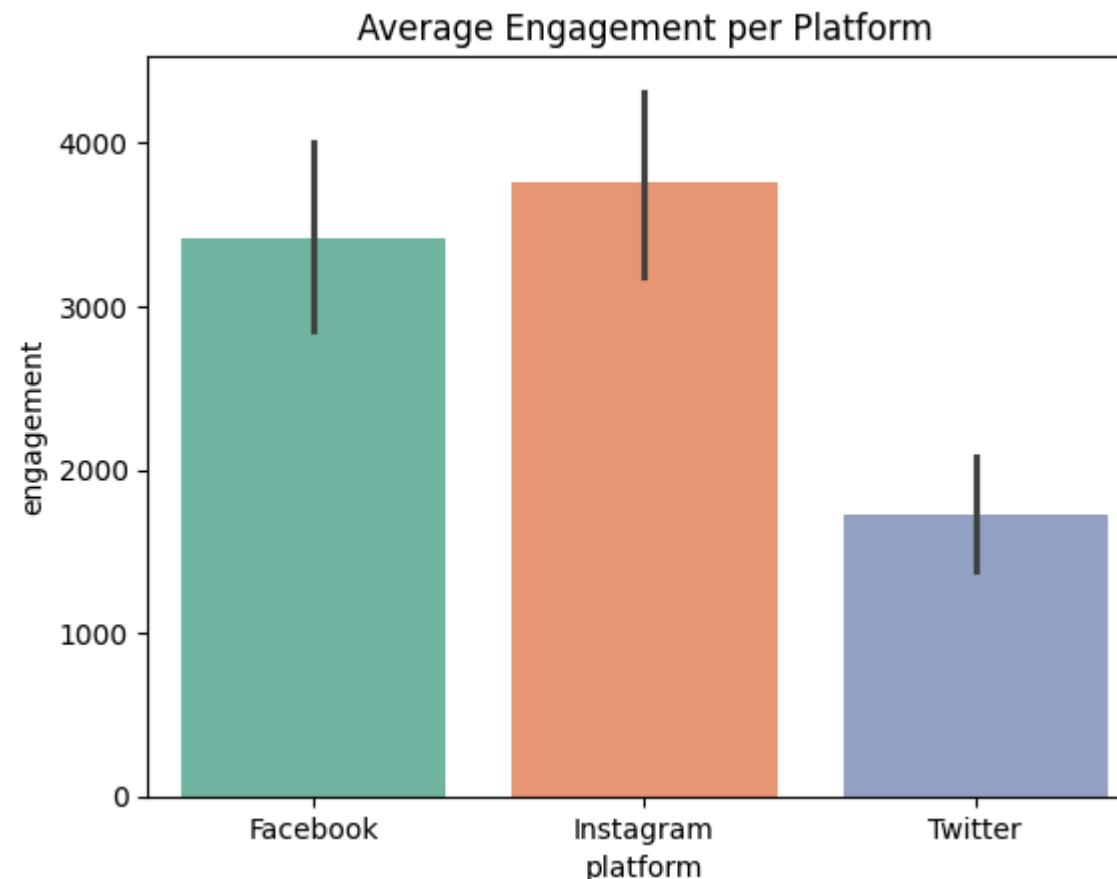
```
In [18]: sns.barplot(
            data=df,
            x="platform",
            y="engagement",
            estimator="mean",
            palette="Set2"  # color palette
)

plt.title("Average Engagement per Platform")
plt.show()
```

```
C:\Users\santo\AppData\Local\Temp\ipykernel_17880\1142403526.py:1: FutureWarning:
```

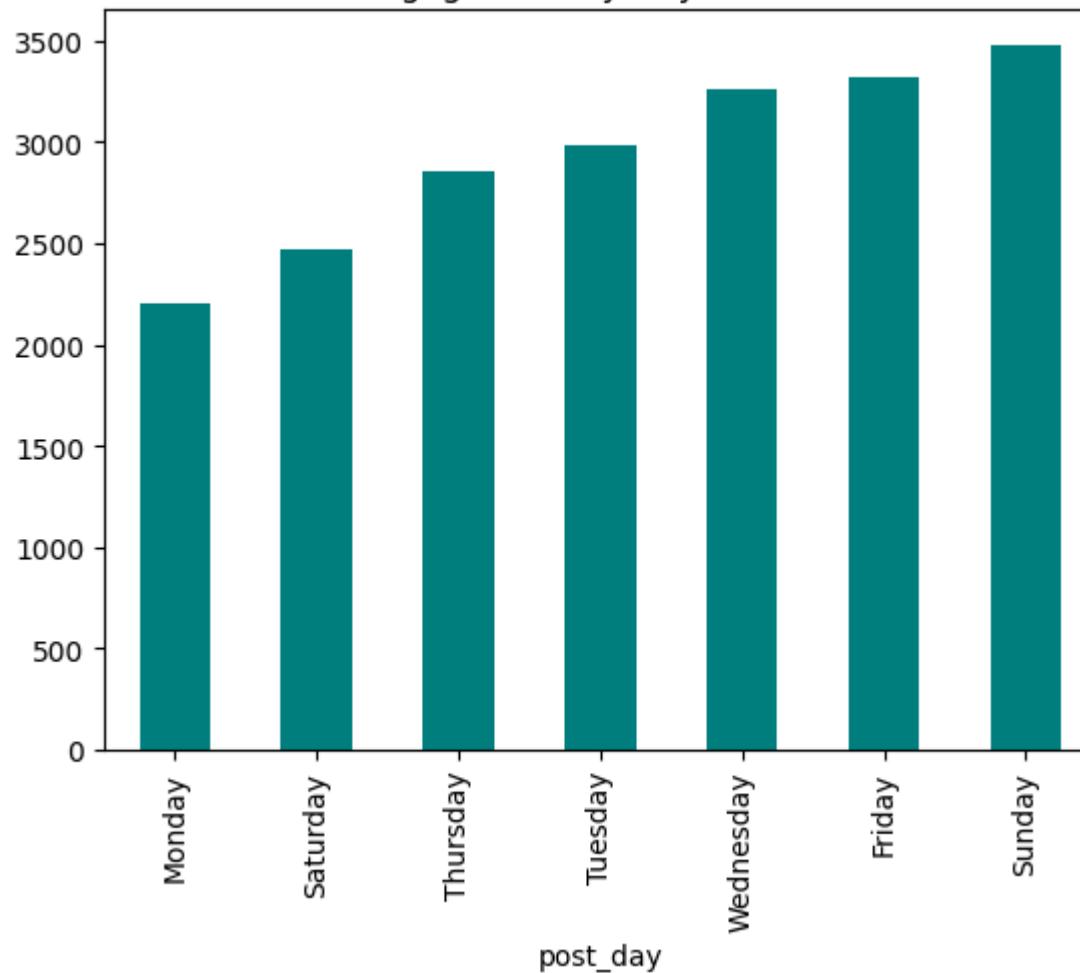
```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.barplot(
```



```
In [19]: engagement_day = df.groupby("post_day")["engagement"].mean().sort_values()  
engagement_day.plot(kind="bar", color="teal", title="Engagement by Day of Week")  
plt.show()
```

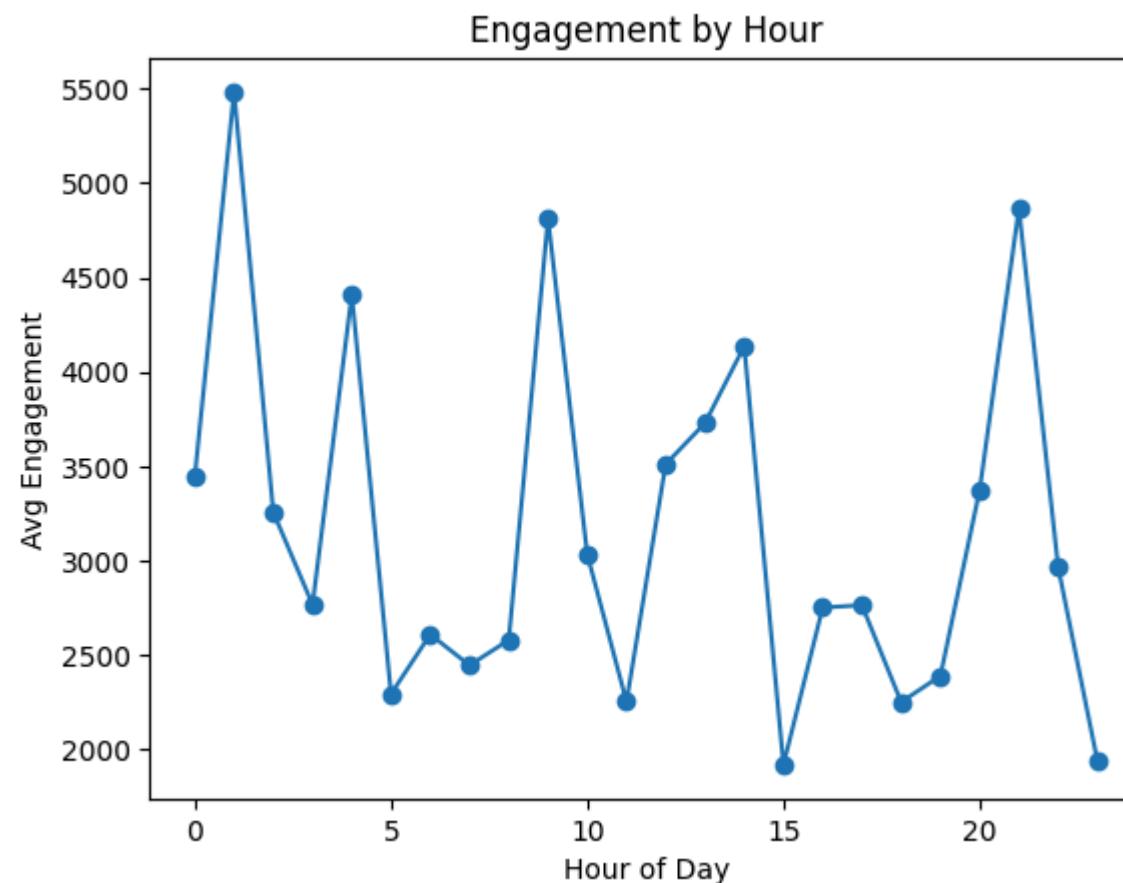
### Engagement by Day of Week



```
In [20]: # Extract hour from post_time
df["post_time"] = pd.to_datetime(df["post_time"])
df["hour"] = df["post_time"].dt.hour

# Engagement by hour
engagement_hour = df.groupby("hour")["engagement"].mean()
engagement_hour.plot(kind="line", marker="o", title="Engagement by Hour")
plt.xlabel("Hour of Day")
```

```
plt.ylabel("Avg Engagement")
plt.show()
```



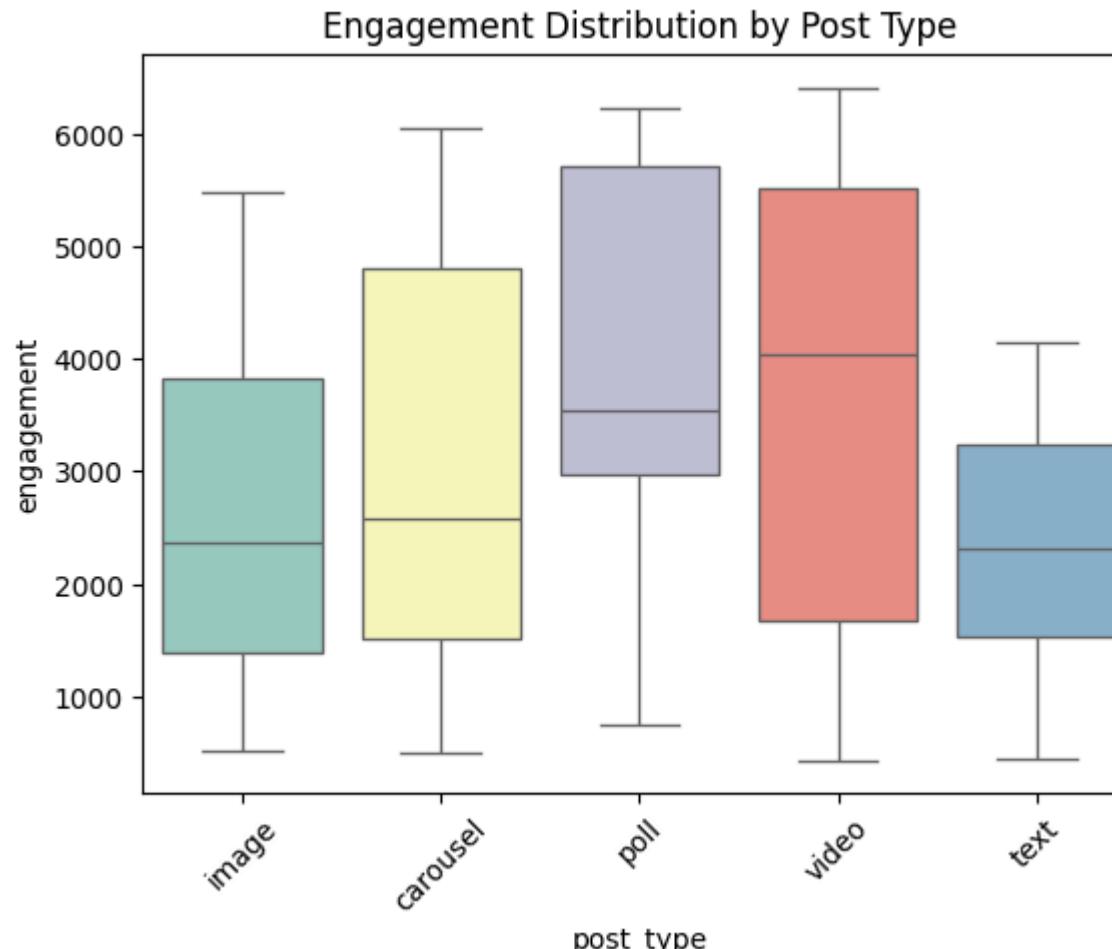
```
In [22]: sns.boxplot(
    data=df,
    x="post_type",
    y="engagement",
    palette="Set3"
)

plt.title("Engagement Distribution by Post Type")
plt.xticks(rotation=45)
plt.show()
```

```
C:\Users\santo\AppData\Local\Temp\ipykernel_17880\2858543308.py:1: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.boxplot(
```

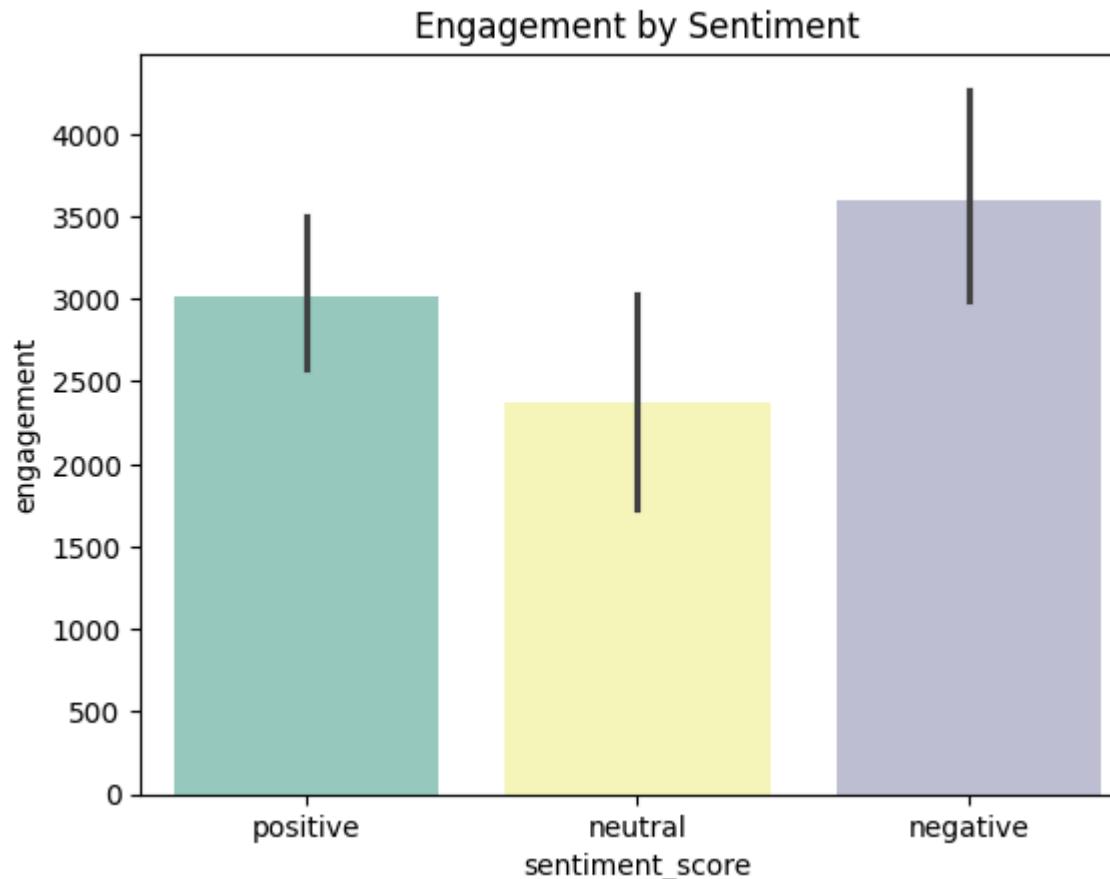


```
In [25]: sns.barplot(data=df, x="sentiment_score", y="engagement", estimator="mean", palette="Set3")
plt.title("Engagement by Sentiment")
plt.show()
```

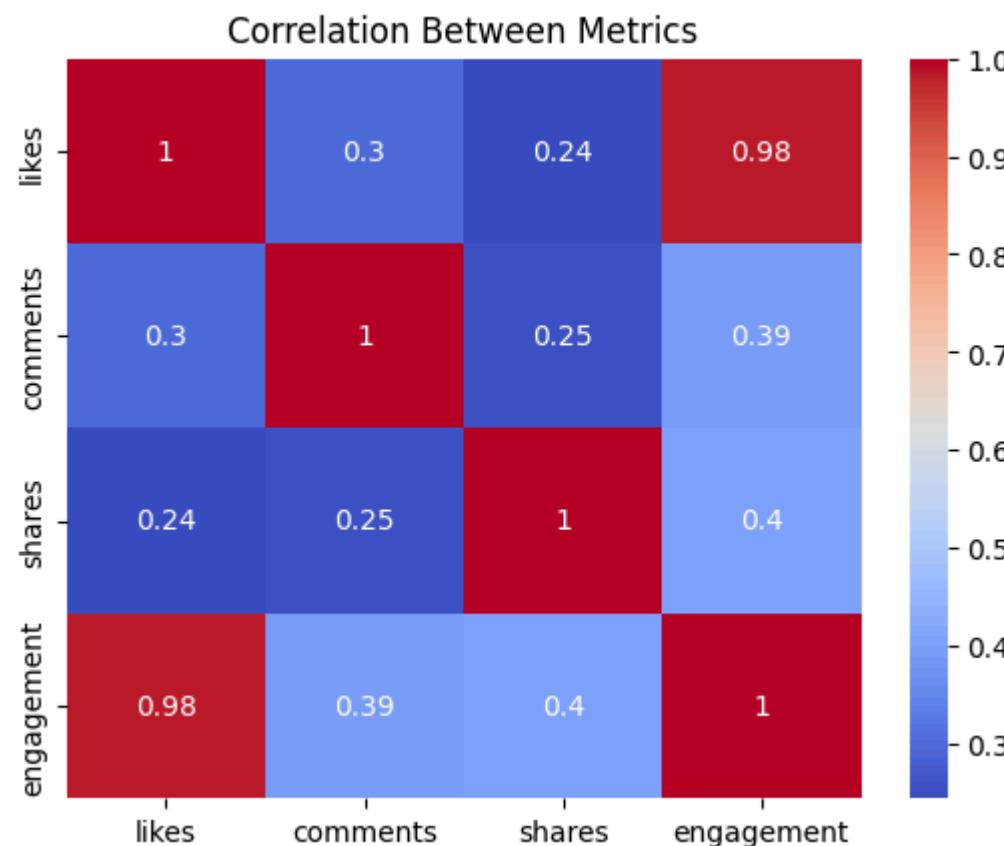
```
C:\Users\santo\AppData\Local\Temp\ipykernel_17880\3590280743.py:1: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(data=df, x="sentiment_score", y="engagement", estimator="mean", palette="Set3")
```



```
In [26]: sns.heatmap(df[["likes", "comments", "shares", "engagement"]].corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Between Metrics")
plt.show()
```



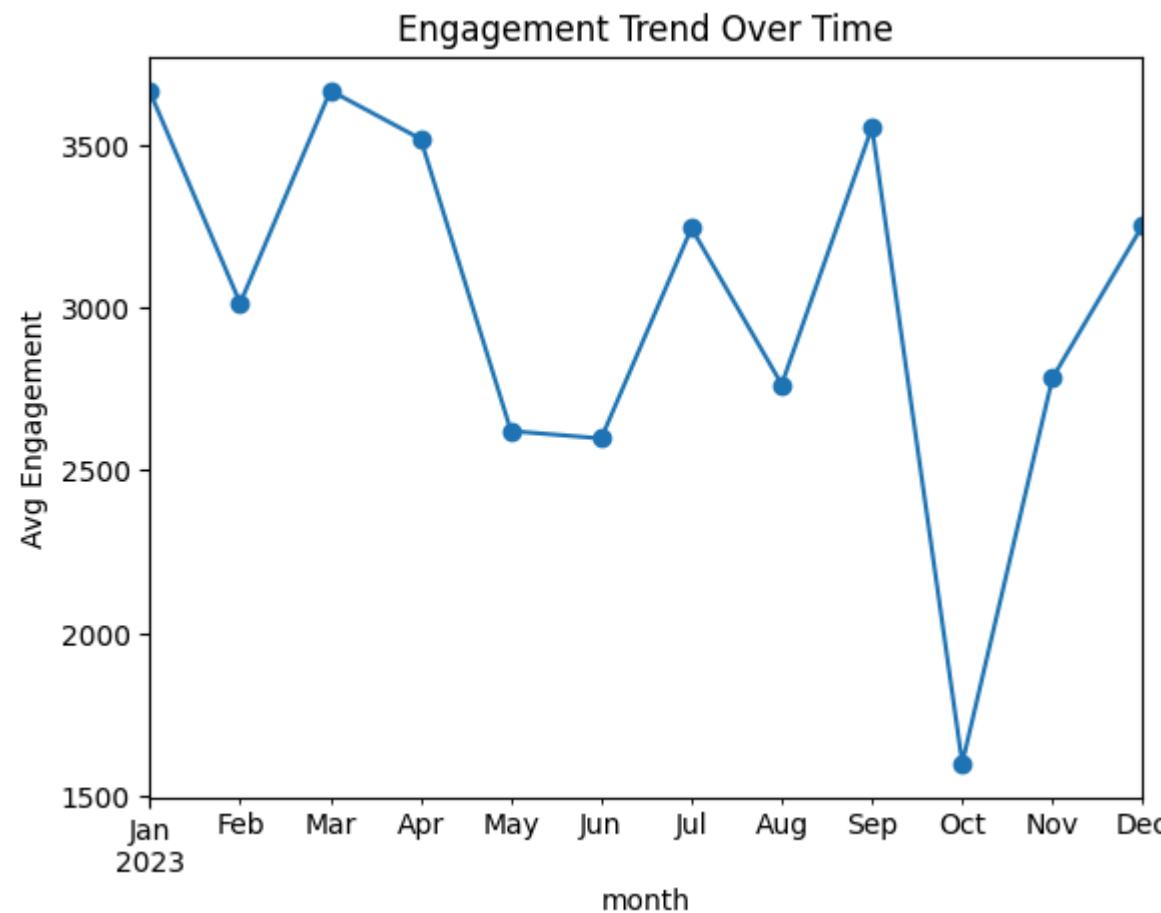
```
In [28]: top_posts = df.nlargest(5, "engagement")[["post_id", "platform", "post_type", "engagement"]]
print(top_posts)
```

	post_id	platform	post_type	engagement
38	39	Facebook	video	6410
30	31	Facebook	poll	6222
73	74	Instagram	video	6198
2	3	Instagram	poll	6051
59	60	Instagram	carousel	6051

```
In [29]: df["month"] = df["post_time"].dt.to_period("M")

engagement_month = df.groupby("month")["engagement"].mean()
engagement_month.plot(kind="line", marker="o", title="Engagement Trend Over Time")
```

```
plt.ylabel("Avg Engagement")
plt.show()
```

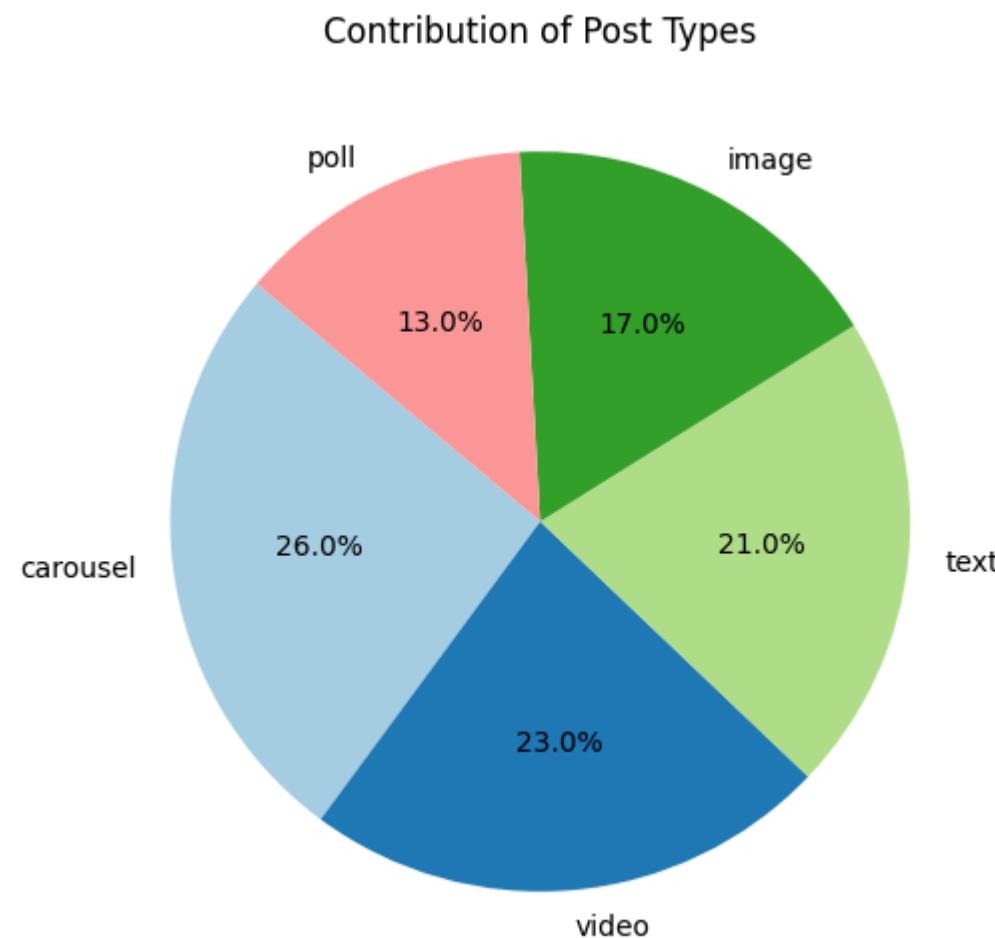


```
In [30]: import matplotlib.pyplot as plt

# Count each post_type
post_type_counts = df["post_type"].value_counts()

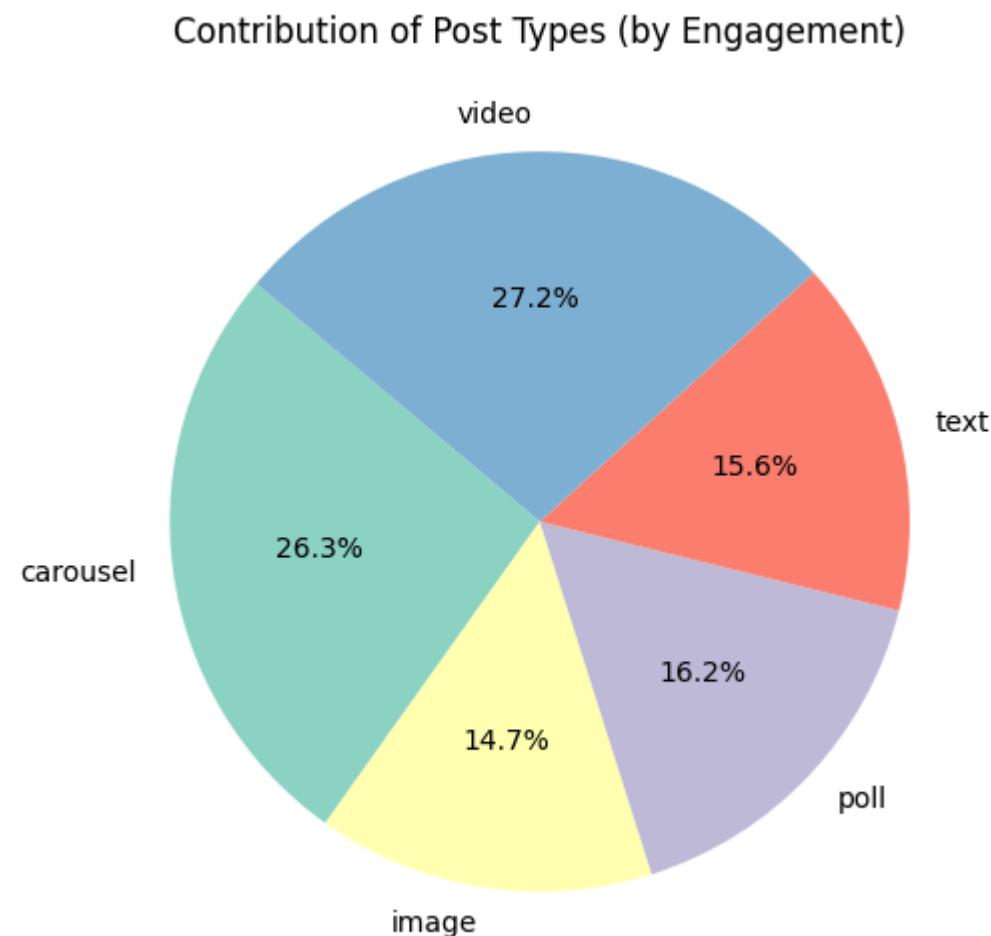
# Pie chart
plt.figure(figsize=(6,6))
plt.pie(
    post_type_counts,
```

```
        labels=post_type_counts.index,  
        autopct='%.1f%%',  
        startangle=140,  
        colors=plt.cm.Paired.colors  
    )  
plt.title("Contribution of Post Types")  
plt.show()
```



```
In [31]: post_type_engagement = df.groupby("post_type")["engagement"].sum()
```

```
plt.figure(figsize=(6,6))
plt.pie(
    post_type_engagement,
    labels=post_type_engagement.index,
    autopct='%.1f%%',
    startangle=140,
    colors=plt.cm.Set3.colors
)
plt.title("Contribution of Post Types (by Engagement)")
plt.show()
```



Summary of the data :

```
from IPython.display import Markdown, display  
summary_md = """"
```

## Key Insights

1. Videos and polls are the best-performing formats across platforms.
2. Instagram drives the highest overall engagement, making it the strongest platform for visual and interactive content.
3. Facebook polls and videos also generate high engagement, showing that interactive content resonates strongly there.
4. Carousels are particularly valuable for shares, amplifying content reach.
5. Twitter's lower performance suggests it may not be the best platform for maximizing engagement with this content mix.
6. Best posting times: 1 AM, 10 AM, and 9 PM are optimal for maximizing engagement.

In [ ]: