

```
In [3]: import pandas as pd  
df=pd.read_csv("Apple_Store_Reviews.csv")  
df
```

Review Statistics							
	Review_ID	App_Name	User_Age	Review_Date	Rating	Review_Text	Likelihood
0	1	Candy Crush Saga	21	2023-01-16	4	Great game, but too many in-game purchases.	Medium
1	2	Spotify	57	2024-02-01	1	Good, but has connection issues sometimes.	Low
2	3	TikTok	33	2023-11-30	5	Awesome app! Best entertainment content.	High
3	4	Audible	40	2023-04-03	5	Great app, but it's a bit pricey.	Medium
4	5	Spotify	44	2023-05-01	1	Good, but has connection issues sometimes.	Low
...							
995	996	Headspace	30	2023-11-15	3	Good, but the premium content is expensive.	Medium
996	997	Duolingo	19	2024-09-27	1	Disappointing. Hard to follow and buggy.	Low
997	998	Duolingo	38	2023-06-07	5	Excellent for learning new skills!	High
998	999	Instagram	52	2024-03-04	4	Great app, but sometimes it lags.	Medium
999	1000	Audible	25	2024-02-20	2	Terrible. Very limited selection of books.	Very Low

1000 rows × 12 columns

```
In [8]: df.columns
```

```
Out[8]: Index(['Review_ID', 'App_Name', 'User_Age', 'Review_Date', 'Rating',
   'Review_Text', 'Likes', 'Device_Type', 'Version_Used', 'Country',
   'Purchase_Amount', 'Category'],
  dtype='object')
```

Calculate the mean, median, and mode of the app ratings in the dataset. Which measure (mean, median, or mode) best represents the central tendency of the ratings?

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

```
Out[9]:
```

	Review_ID	User_Age	Rating	Likes	Purchase_Amount
count	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000
mean	500.500000	39.211000	2.869000	44.776000	5.361120
std	288.819436	11.908917	1.467649	28.685444	5.755652
min	1.000000	18.000000	1.000000	0.000000	0.000000
25%	250.750000	30.000000	1.000000	17.000000	0.000000
50%	500.500000	39.000000	3.000000	42.500000	4.995000
75%	750.250000	49.000000	4.000000	71.000000	10.192500
max	1000.000000	60.000000	5.000000	100.000000	19.970000

```
In [11]: mean_ratings = df["Rating"].mean()
mean_ratings
```

```
Out[11]: np.float64(2.869)
```

```
In [12]: median_ratings = df["Rating"].median()
median_ratings
```

```
Out[12]: np.float64(3.0)
```

```
In [14]: mode_ratings = df["Rating"].mode()[0]
mode_ratings
```

```
Out[14]: np.int64(1)
```

```
In [ ]: #The median rating best represents the central
#tendency of app ratings because it is less affected by extreme values and s
```

Find the range and interquartile range (IQR) of the Purchase_Amount in the dataset. How do these values help in understanding the spread of the data?

```
In [15]: #range  
range_purchase_Amount=df["Purchase_Amount"].max()-df["Purchase_Amount"].min()  
range_purchase_Amount
```

```
Out[15]: np.float64(19.97)
```

```
In [21]: #IQR=q3-q1  
Q1=df["Purchase_Amount"].quantile(0.25)  
Q3=df["Purchase_Amount"].quantile(0.75)  
Q1,Q3  
#Q1 is zero mean 25% of users does not purchase anything and 75% of purchases
```

```
Out[21]: (np.float64(0.0), np.float64(10.19249999999999))
```

```
In [22]: IQR = Q3-Q1  
print("The interquartile range (IQR) of the Purchase_Amount is approximately")
```

The interquartile range (IQR) of the Purchase_Amount is approximately 10.19, indicating that the middle 50% of purchases lie between 0 and 10.19. Since the first quartile is 0, this suggests a significant proportion of users made no purchase, leading to a right-skewed distribution. 10.19249999999999

Calculate the variance and standard deviation for the number of likes received on reviews. What does the standard deviation indicate about the spread of the data?

```
In [29]: variance_like = df["Likes"].var()  
std_like=df["Likes"].std()  
print("This is Variance (sigma)^2: ",variance_like,"This is std (sigma): ",s  
print()  
print("Although the average number of likes per review is low (2.86), the ve
```

```
This is Variance (sigma)^2:  822.8546786786787 This is std (sigma):  28.6854  
43672334557
```

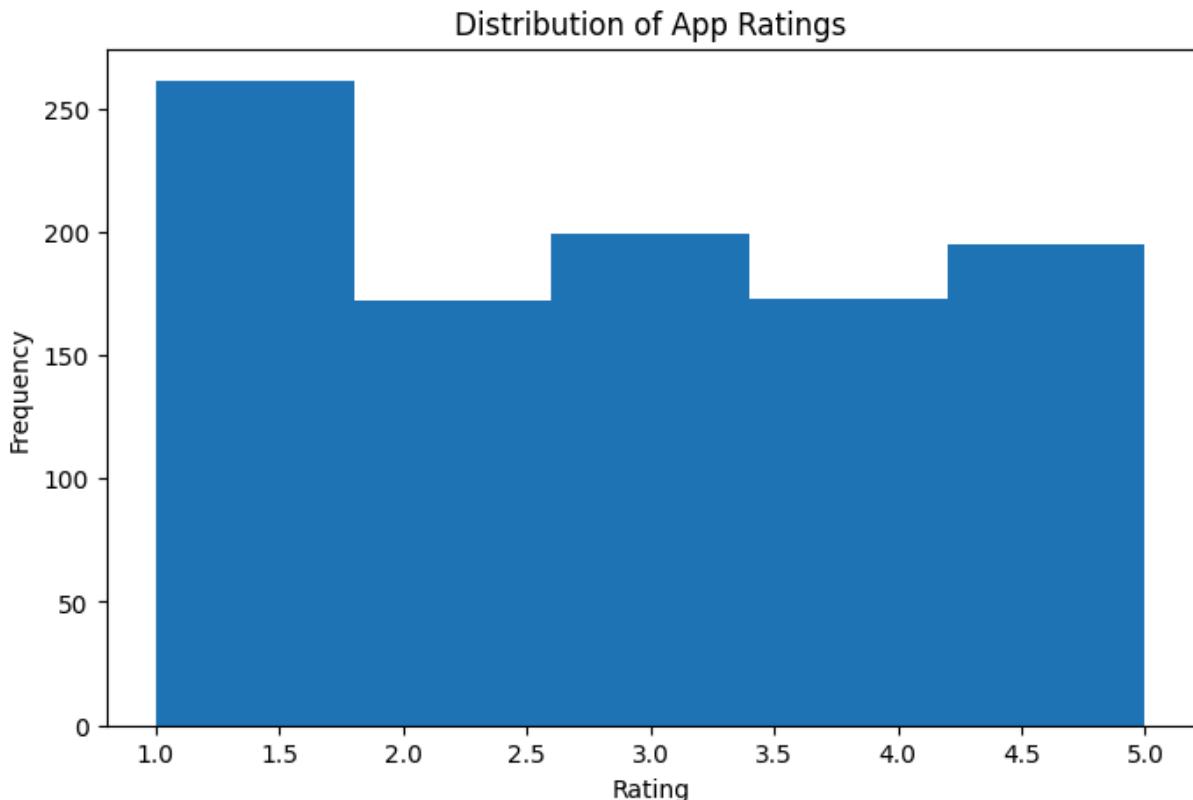
Although the average number of likes per review is low (2.86), the very high standard deviation (28.9) indicates extreme variability in engagement. This suggests that while most reviews receive minimal likes, a small number of reviews gain disproportionately high attention, leading to a highly right-skewed distribution.

Determine the correlation between the likes and the rating given. Is there a positive, negative, or no correlation between these variables?

```
In [33]: corr = df["Likes"].corr(df["Rating"])  
corr  
print("corr values is 0.84 after cal. which is close to 1 it means this is (+)ve co  
relation. This indicates strong Linear relationship")
```

Plot the distribution of the app ratings. Is the distribution positively or negatively skewed? What does this indicate about user satisfaction?

```
In [39]: import matplotlib.pyplot as plt  
  
plt.figure(figsize=(8,5))  
plt.hist(df["Rating"], bins=5) #rating from 1 to 5  
plt.xlabel("Rating")  
plt.ylabel("Frequency")  
plt.title("Distribution of App Ratings")  
plt.show()  
print("""The histogram shows that app ratings are more concentrated toward h  
This results in a negatively skewed distribution, indicating that most users  
while a smaller proportion report dissatisfaction.""")
```



The histogram shows that app ratings are more concentrated toward higher values (3 to 5), with fewer low ratings (1 to 2).

This results in a negatively skewed distribution, indicating that most users are generally satisfied with the apps, while a smaller proportion report dissatisfaction.

```
In [40]: # Perform a hypothesis test to determine if the average rating for Instagram  
# step-1 Null Hypothesis [h0 , mew(instagram) <= mew(whatsapp)]  
# step-2 Alternative Hypothesis [h1 , mew(instagram) > mew(whatsapp)]  
insta_rating = df[df["App_Name"]=="Instagram"]["Rating"]  
whatsapp_rating = df[df["App_Name"]=="Whatsapp"]["Rating"]
```

```
In [41]: pip install scipy
```

```
Collecting scipy
```

```
  Downloading scipy-1.17.0-cp314-cp314-win_amd64.whl.metadata (60 kB)
Requirement already satisfied: numpy<2.7,>=1.26.4 in c:\users\dell\appdata\local\programs\python\python314\lib\site-packages (from scipy) (2.3.3)
  Downloading scipy-1.17.0-cp314-cp314-win_amd64.whl (37.1 MB)
----- 0.0/37.1 MB ? eta ------
----- 1.0/37.1 MB 6.2 MB/s eta 0:00:06
----- 1.8/37.1 MB 4.8 MB/s eta 0:00:08
----- 2.6/37.1 MB 4.4 MB/s eta 0:00:08
----- 3.4/37.1 MB 4.3 MB/s eta 0:00:08
----- 4.5/37.1 MB 4.2 MB/s eta 0:00:08
----- 4.7/37.1 MB 4.0 MB/s eta 0:00:09
----- 5.5/37.1 MB 3.8 MB/s eta 0:00:09
----- 6.0/37.1 MB 3.8 MB/s eta 0:00:09
----- 6.8/37.1 MB 3.5 MB/s eta 0:00:09
----- 7.6/37.1 MB 3.6 MB/s eta 0:00:09
----- 8.4/37.1 MB 3.6 MB/s eta 0:00:08
----- 9.2/37.1 MB 3.6 MB/s eta 0:00:08
----- 10.2/37.1 MB 3.6 MB/s eta 0:00:0
8----- 10.7/37.1 MB 3.6 MB/s eta 0:00:0
8----- 11.5/37.1 MB 3.5 MB/s eta 0:00:0
8----- 12.3/37.1 MB 3.6 MB/s eta 0:00:0
7----- 13.1/37.1 MB 3.6 MB/s eta 0:00:0
7----- 13.9/37.1 MB 3.6 MB/s eta 0:00:0
7----- 14.9/37.1 MB 3.6 MB/s eta 0:00:0
7----- 15.7/37.1 MB 3.6 MB/s eta 0:00:0
6----- 16.5/37.1 MB 3.6 MB/s eta 0:00:0
6----- 17.3/37.1 MB 3.6 MB/s eta 0:00:0
6----- 18.4/37.1 MB 3.6 MB/s eta 0:00:0
6----- 19.1/37.1 MB 3.6 MB/s eta 0:00:0
5----- 19.7/37.1 MB 3.6 MB/s eta 0:00:0
5----- 20.2/37.1 MB 3.6 MB/s eta 0:00:0
5----- 20.4/37.1 MB 3.5 MB/s eta 0:00:0
5----- 21.0/37.1 MB 3.5 MB/s eta 0:00:0
5----- 21.2/37.1 MB 3.4 MB/s eta 0:00:0
5----- 21.8/37.1 MB 3.4 MB/s eta 0:00:0
5----- 22.3/37.1 MB 3.3 MB/s eta 0:00:0
```

```
----- 23.3/37.1 MB 3.3 MB/s eta 0:00:0
5
----- 24.1/37.1 MB 3.3 MB/s eta 0:00:0
4
----- 24.9/37.1 MB 3.3 MB/s eta 0:00:0
4
----- 25.7/37.1 MB 3.4 MB/s eta 0:00:0
4
----- 26.5/37.1 MB 3.4 MB/s eta 0:00:0
4
----- 27.0/37.1 MB 3.3 MB/s eta 0:00:0
4
----- 27.3/37.1 MB 3.3 MB/s eta 0:00:0
3
----- 27.8/37.1 MB 3.3 MB/s eta 0:00:0
3
----- 28.6/37.1 MB 3.3 MB/s eta 0:00:0
3
----- 29.6/37.1 MB 3.3 MB/s eta 0:00:0
3
----- 30.7/37.1 MB 3.3 MB/s eta 0:00:0
2
----- 31.5/37.1 MB 3.3 MB/s eta 0:00:0
2
----- 32.0/37.1 MB 3.3 MB/s eta 0:00:0
2
----- 32.5/37.1 MB 3.3 MB/s eta 0:00:0
2
----- 32.8/37.1 MB 3.2 MB/s eta 0:00:0
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----- 33.6/37.1 MB 3.2 MB/s eta 0:00:0
2
----- 34.3/37.1 MB 3.2 MB/s eta 0:00:0
1
----- 35.1/37.1 MB 3.3 MB/s eta 0:00:0
1
----- 35.9/37.1 MB 3.3 MB/s eta 0:00:0
1
----- 36.7/37.1 MB 3.3 MB/s eta 0:00:0
1
----- 37.1/37.1 MB 3.2 MB/s 0:00:11
Installing collected packages: scipy
Successfully installed scipy-1.17.0
Note: you may need to restart the kernel to use updated packages.
[notice] A new release of pip is available: 25.2 -> 26.0.1
[notice] To update, run: python.exe -m pip install --upgrade pip
```

In [43]: `from scipy.stats import ttest_ind`

```
t_stat, p_value = ttest_ind(
    insta_rating,
    whatsapp_rating,
    equal_var=False # Welch's t-test (safe choice)
)
```

```
t_stat, p_value
```

```
C:\Users\DELL\AppData\Local\Temp\ipykernel_2520\504490773.py:3: SmallSampleWarning: One or more sample arguments is too small; all returned values will be NaN. See documentation for sample size requirements.  
    t_stat, p_value = ttest_ind(
```

```
Out[43]: (np.float64(nan), np.float64(nan))
```

```
In [44]: one_tailed_p = p_value / 2  
one_tailed_p
```

```
Out[44]: np.float64(nan)
```

```
In [45]: alpha = 0.05
```

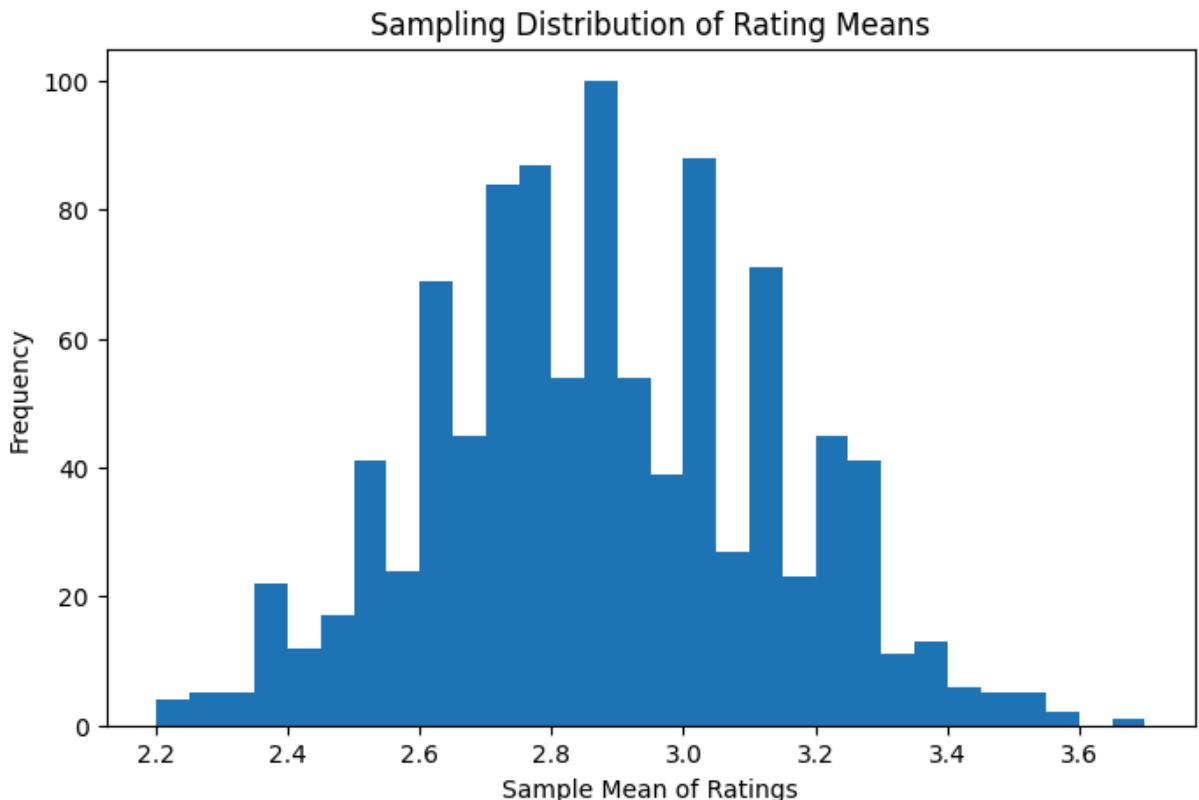
```
if one_tailed_p < alpha:  
    print("Reject H0: Instagram average rating is significantly higher than  
else:  
    print("Fail to reject H0: No significant difference found.")
```

```
Fail to reject H0: No significant difference found.
```

Take random samples of ratings from the dataset and calculate their means. Create a sampling distribution and explain how this relates to the Central Limit Theorem.

```
In [46]: import numpy as np  
import matplotlib.pyplot as plt  
  
# Parameters  
sample_size = 30          # size of each sample  
num_samples = 1000         # number of samples  
  
sample_means = []  
  
for _ in range(num_samples):  
    sample = df["Rating"].sample(n=sample_size, replace=True)  
    sample_means.append(sample.mean())
```

```
In [47]: plt.figure(figsize=(8,5))  
plt.hist(sample_means, bins=30)  
plt.xlabel("Sample Mean of Ratings")  
plt.ylabel("Frequency")  
plt.title("Sampling Distribution of Rating Means")  
plt.show()
```



```
In [48]: print("""Sampling distribution will look:
```

1. Symmetric
2. Bell-shaped
3. Approximately normal

```
✓ This is Central Limit Theorem in action""")
```

Sampling distribution will look:

1. Symmetric
2. Bell-shaped
3. Approximately normal

✓ This is Central Limit Theorem in action

```
In [ ]:
```