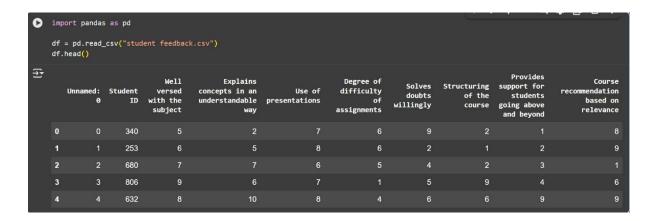
1. Importing the Data into Google Colaboratory

import pandas as pd

df = pd.read_csv("student feedback.csv")
df.head()

Result:



#Check column names and data info

print(df.info())

Result:

Explains concepts in an understandable way Use of presentations \

0	2	7
1	5	8
2	7	6
3	6	7
4	10	8

Degree of difficulty of assignments Solves doubts willingly \

0	6	9
1	6	2
2	5	4
3	1	5
4	4	6

Structuring of the course \

0	2
1	1
2	2

3	9	
4	6	
I	Provides support for students going ab	pove and beyond \
0	1	•
1	2	
2	3	
3	4	
4	9	
(Course recommendation based on rele	vance
0	8	
1	9	
2	1	
3	6	
4	9	
<c< td=""><td>lass 'pandas.core.frame.DataFrame'></td><td></td></c<>	lass 'pandas.core.frame.DataFrame'>	
Ra	ngeIndex: 1001 entries, 0 to 1000	
Da	ta columns (total 10 columns):	
#	Column	Non-Null Count Dtype
	Unnamed: 0	1001 non-null int64
1	Student ID	1001 non-null int64
	Well versed with the subject	1001 non-null int64
3	Explains concepts in an understanda	-
	Use of presentations	1001 non-null int64
5	Degree of difficulty of assignments	1001 non-null int64
6	Solves doubts willingly	1001 non-null int64
7	Structuring of the course	1001 non-null int64
8		above and beyond 1001 non-null int64
9	Course recommendation based on re	elevance 1001 non-null int64
-	rpes: int64(10)	
	emory usage: 78.3 KB	
No	ne	

2. Understanding the Data

Check number of rows, columns

print("Shape:", df.shape)
#Checking for number of rows and columns

print("Columns:", df.columns.tolist()) # To see the column names

print(df.info()) #Basic information like data types and nulls

print(df.describe()) # Quick stats for numbers like range

Result:

Shape: (1001, 10)

Columns: ['Unnamed: 0', 'Student ID', 'Well versed with the subject', 'Explains concepts in an understandable way', 'Use of presentations', 'Degree of difficulty of assignments', 'Solves doubts willingly', 'Structuring of the course', 'Provides support for students going above and beyond', 'Course recommendation based on relevance'] <class 'pandas.core.frame.DataFrame'>

RangeIndex: 1001 entries, 0 to 1000 Data columns (total 10 columns):

Column Non-Null Count Dtype

--- ----- -----

0 Unnamed: 0 1001 non-null int64 1 Student ID 1001 non-null int64

2 Well versed with the subject 1001 non-null int64

3 Explains concepts in an understandable way 1001 non-null int64

4 Use of presentations 1001 non-null int64

5 Degree of difficulty of assignments 1001 non-null int64

6 Solves doubts willingly 1001 non-null int64 7 Structuring of the course 1001 non-null int64

8 Provides support for students going above and beyond 1001 non-null int64

9 Course recommendation based on relevance 1001 non-null int64

dtypes: int64(10)

memory usage: 78.3 KB

None

Unnamed: 0 Student ID Well versed with the subject \

 count 1001.000000 1001.000000
 1001.000000

 mean 500.000000 500.000000
 7.497502

 std 289.108111 289.108111
 1.692998

 min 0.000000 0.000000
 5.000000

25%	250.000000	250.000000	6.000000
50%	500.000000	500.000000	8.000000
75%	750.000000	750.000000	9.000000
max	1000.000000	1000.000000	10.000000

Explains concepts in an understandable way Use of presentations \

count	1001.000000	1001.000000
mean	6.081918	5.942058
std	2.597168	1.415853
min	2.000000	4.000000
25%	4.000000	5.000000
50%	6.000000	6.000000
75%	8.000000	7.000000
max	10.000000	8.000000

count	1001.000000	1001.000000
mean	5.430569	5.474525
std	2.869046	2.874648
min	1.000000	1.000000
25%	3.000000	3.000000
50%	5.000000	6.000000
75%	8.000000	8.000000
max	10.000000	10.000000

Structuring of the course \

	_
count	1001.000000
mean	5.636364
std	2.920212
min	1.000000
25%	3.000000
50%	6.000000
75%	8.000000
max	10.000000

Provides support for students going above and beyond \

count	1001.000000
mean	5.662338
std	2.891690
min	1.000000
25%	3.000000
50%	6.000000
75%	8.000000
max	10.000000

Course recommendation based on relevance

count	1001.000000
mean	5.598402
std	2.886617
min	1.000000
25%	3.000000
50%	6.000000
75%	8.000000
max	10.000000

3. Cleaning the Dataset (Removing the unnecessary columns)

```
df_clean = df.drop(columns=['Unnamed: 0', 'Student ID'])
# Dropping unwanted columns
df_clean.head()
```

Result:

	df_cl	lean = df.dr	op(columns=['Unnamed:	0', 'Student ID'	(1)				
	df_cl	lean.head()							
3		Well versed with the subject	Explains concepts in an understandable way	Use of presentations	Degree of difficulty of assignments	Solves doubts willingly	Structuring of the course	Provides support for students going above and beyond	Course recommendation based on relevance
	0	5	2	7	6	9	2		8
	1	6	5	8	6	2		2	9
	2			6		4	2	3	
	3	9	6	7		5	9	4	6
	4	8	10	8	4	6	6	9	9

4. Finding the average rating for each question.

```
avg_ratings = df_clean.mean()  # Average rating for each question

print("Average Ratings:")

print(avg_ratings)  # Printing average ratings
```

Result:

Average Ratings:

Well versed with the subject 7.497502

Explains concepts in an understandable way 6.081918

Use of presentations 5.942058

Degree of difficulty of assignments 5.430569
Solves doubts willingly 5.474525
Structuring of the course 5.636364

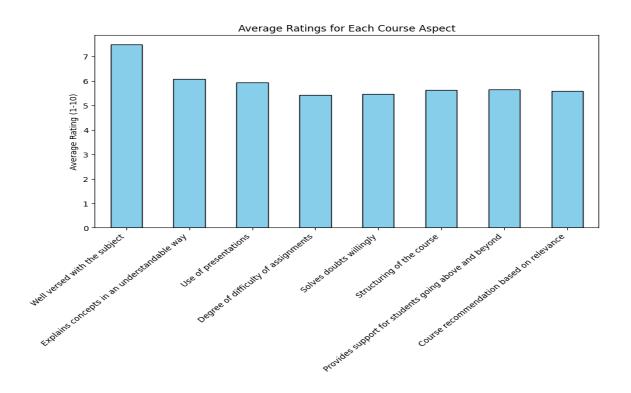
Provides support for students going above and beyond 5.662338 Course recommendation based on relevance 5.598402

dtype: float64

5. Visualize the Averages

import matplotlib.pyplot as plt

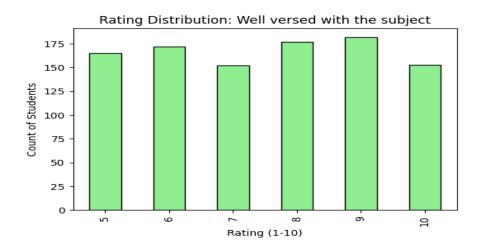
```
avg_ratings.plot(kind='bar', figsize=(10,5), color='skyblue', edgecolor='black') plt.title("Average Ratings for Each Course Aspect") plt.ylabel("Average Rating (1-10)") plt.xticks(rotation=45, ha='right') plt.show()
```

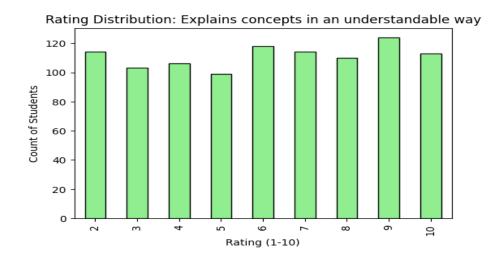


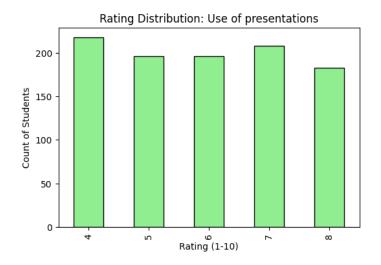
6. Distribution of Ranges

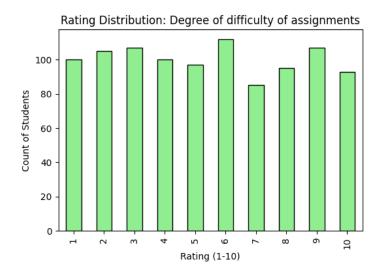
import matplotlib.pyplot as plt

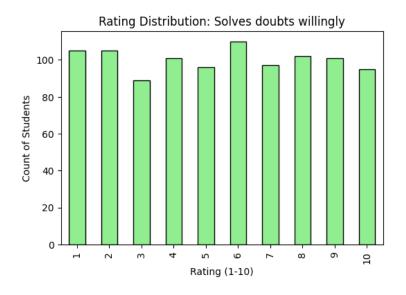
```
for col in df_clean.columns:
    plt.figure(figsize=(6,4))
    df_clean[col].value_counts().sort_index().plot(kind='bar', color='lightgreen',
edgecolor='black')
    plt.title(f''Rating Distribution: {col}")
    plt.xlabel("Rating (1-10)")
    plt.ylabel("Count of Students")
    plt.show()
```

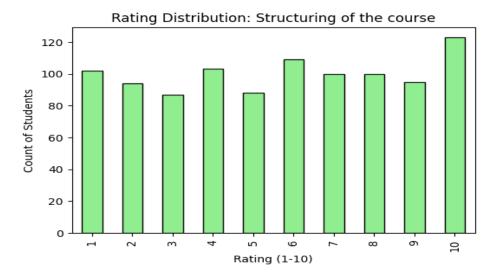




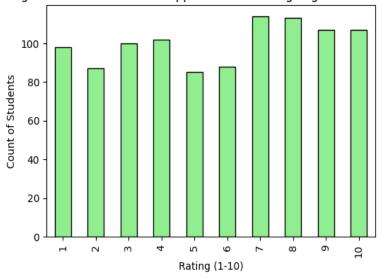




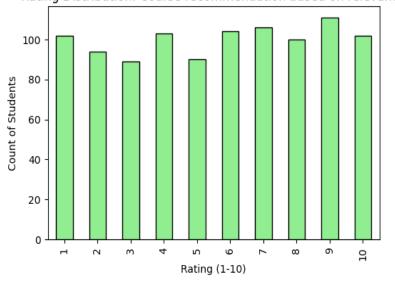




Rating Distribution: Provides support for students going above and beyond



Rating Distribution: Course recommendation based on relevance

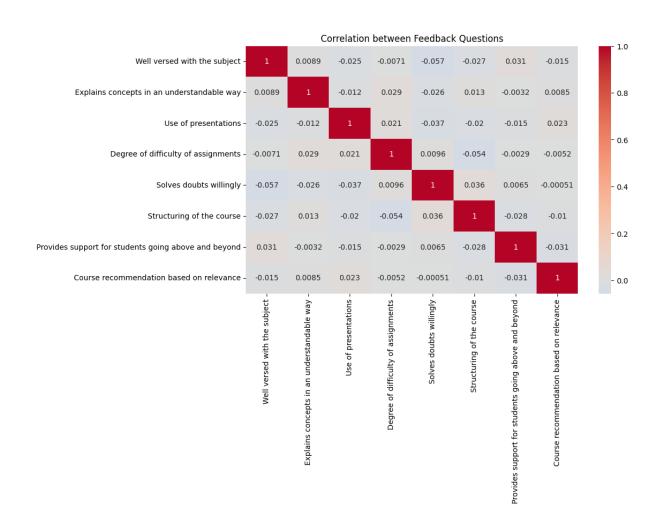


7. Correlation Analysis

import seaborn as sns import matplotlib.pyplot as plt

```
# Compute correlation matrix (ignoring ID columns)
corr = df.drop(columns=["Unnamed: 0", "Student ID"]).corr()
```

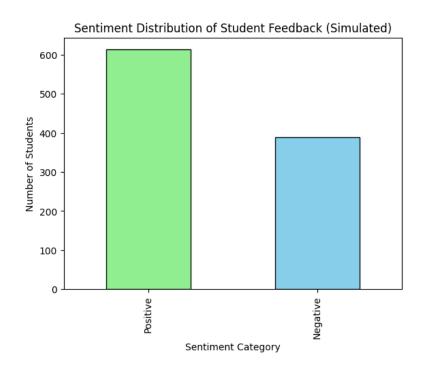
plt.figure(figsize=(10, 6)) # Plot the Heatmap sns.heatmap(corr, annot=True, cmap="coolwarm", center=0) plt.title("Correlation between Feedback Questions") plt.show()



8. Sentiment Analysis (NLP)

```
import pandas as pd
import numpy as np
from textblob import TextBlob
import matplotlib.pyplot as plt
from wordcloud import WordCloud
# Load your dataset again
df = pd.read_csv("/feedback.csv")
# Step 1: Create synthetic comments based on ratings
# We'll map overall Course recommendation score to sentiment-like comments
def generate comment(score):
  if score \geq= 8:
    return np.random.choice([
       "The event was excellent and very useful",
       "Loved the session, very engaging and informative",
       "Great experience, learned a lot"
    ])
  elif score \geq = 5:
    return np.random.choice([
       "It was okay, but could be improved",
       "The session was average, some parts were useful",
       "Not bad, but expected a bit more"
    ])
  else:
    return np.random.choice([
       "I did not like it, too boring",
       "The workshop was confusing and not helpful",
       "Poorly conducted, needs improvement"
    1)
df['Comments'] = df['Course recommendation based on
relevance'].apply(generate comment)
# Step 2: Perform sentiment analysis with TextBlob
df['sentiment score'] = df['Comments'].apply(lambda x:
TextBlob(str(x)).sentiment.polarity)
df['sentiment label'] = df['sentiment score'].apply(
  lambda x: 'Positive' if x > 0 else ('Negative' if x < 0 else 'Neutral')
)
```

```
# Step 3: Visualize sentiment distribution
df['sentiment label'].value counts().plot(
  kind='bar', color=['lightgreen', 'skyblue', 'salmon'], edgecolor='black'
)
plt.title("Sentiment Distribution of Student Feedback (Simulated)")
plt.ylabel("Number of Students")
plt.xlabel("Sentiment Category")
plt.show()
# Step 4: Word cloud
text = " ".join(comment for comment in df['Comments'])
wordcloud = WordCloud(width=800, height=400,
background color='white').generate(text)
plt.figure(figsize=(10,5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.title("Most Common Words in Feedback Comments (Simulated)")
plt.show()
```





9. Recommendations based on Findings

```
import pandas as pd
# Load your dataset
df = pd.read csv("feedback.csv")
# Remove unwanted ID columns
df clean = df.drop(columns=['Unnamed: 0', 'Student ID'])
# Calculate average ratings
avg ratings = df clean.mean()
# Recommendation system function
def generate recommendations(avg ratings):
  recs = []
  high threshold = 7.5
  low threshold = 6.0
  if avg ratings['Well versed with the subject'] >= high threshold:
    recs.append("Maintain instructors' subject expertise and encourage continuous
learning.")
  else:
    recs.append("Improve the subject knowledge of instructors through training.")
  if avg_ratings['Use of presentations'] < low_threshold:
    recs.append("Enhance presentation quality using better visuals and interactive
materials.")
```

```
if avg_ratings['Degree of difficulty of assignments'] < low_threshold:
    recs.append("Adjust assignment difficulty to balance challenge and feasibility.")
  if avg ratings['Solves doubts willingly'] < low threshold:
    recs.append("Increase availability of instructors for doubt solving sessions.")
  if avg_ratings['Structuring of the course'] < low_threshold:
    recs.append("Revisit course structure, pacing, and flow to improve student
experience.")
  if avg_ratings['Provides support for students going above and beyond'] <
low threshold:
    recs.append("Offer additional support for advanced and struggling students.")
  if avg ratings['Course recommendation based on relevance'] < low threshold:
    recs.append("Ensure course content is relevant and updated to match student and
industry needs.")
  return recs
# Generate and print recommendations
recommendations = generate recommendations(avg ratings)
print("Recommendations based on feedback ratings:\n")
for i, rec in enumerate(recommendations, 1):
  print(f"{i}. {rec}")
```

Result:

Recommendations based on feedback ratings:

- 1. Improve the subject knowledge of instructors through training.
- 2. Enhance presentation quality using better visuals and interactive materials.
- 3. Adjust assignment difficulty to balance challenge and feasibility.
- 4. Increase availability of instructors for doubt solving sessions.
- 5. Revisit course structure, pacing, and flow to improve student experience.
- 6. Offer additional support for advanced and struggling students.
- 7. Ensure course content is relevant and updated to match student and industry needs.