Task

Analyze the student feedback data from "/student_feedback.csv" to identify satisfaction levels and areas for improvement. Calculate the average rating per question (excluding the first two columns), identify strengths and weaknesses, and visualize the results using a bar chart of average ratings by category, a boxplot of rating variations, a heatmap of correlations between feedback aspects, and a pie chart of the average score for "Course recommendation".

Load data

Subtask:

Load the student feedback data from the provided CSV file into a pandas DataFrame.

Reasoning: Import pandas and load the data into a DataFrame, then display the head and info of the DataFrame to inspect it.

```
import pandas as pd

df = pd.read_csv("/student_feedback.csv")
display(df.head())
display(df.info())
```



	Unnamed: 0	Student ID	Well versed with the subject	Explains concepts in an understandable way		Degree of difficulty of assignments	Solves doubts willingly	Str
)	0	340	5	2	7	6	9	
	1	253	6	5	8	6	2	
!	2	680	7	7	6	5	4	
}	3	806	9	6	7	1	5	
ļ	4	632	8	10	8	4	6	
<pre>:lass 'pandas.core.frame.DataFrame'> ingeIndex: 1001 entries, 0 to 1000 ita columns (total 10 columns): ! Column Non-Null Count</pre>						Dtype 		
)	Unnamed: 0					001 non-null	int64	
. Student ID						001 non-null	int64	
!	! Well versed with the subject					001 non-null	int64	
j	Explains concepts in an understandable way Use of presentations					001 non-null 001 non-null	int64 int64	
;	Degree of difficulty of assignments					001 non-null	int64	
;	Solves doubts willingly					001 non-null	int64	
,	Structuring of the course					001 non-null	int64	
}						001 non-null	int64	
)	Course recommendation based on relevance					001 non-null	int64	
:)	/pes: int64	1(10)						
'n	nory usage:	78.3 KB						

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Analyze ratings

Subtask:

Calculate and analyze the average ratings for each question, excluding the first two columns. Identify strengths and areas for improvement based on these ratings.

Reasoning: Calculate the average rating for each feedback question and identify strengths and weaknesses based on these averages.

```
feedback_ratings = df.iloc[:, 2:]
average_ratings = feedback_ratings.mean()
print("Average ratings for each feedback aspect:")
print(average ratings)
print("\nStrengths (highest average ratings):")
print(average ratings.nlargest(3))
print("\nAreas for improvement (lowest average ratings):")
print(average_ratings.nsmallest(3))
→ Average ratings for each feedback aspect:
     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
     Strengths (highest average ratings):
     Well versed with the subject
                                                   7.497502
     Explains concepts in an understandable way
                                                   6.081918
     Use of presentations
                                                   5.942058
     dtype: float64
     Areas for improvement (lowest average ratings):
     Degree of difficulty of assignments
                                                 5.430569
     Solves doubts willingly
                                                 5.474525
     Course recommendation based on relevance
                                                 5.598402
     dtype: float64
```

Visualize rating analysis

Subtask:

Create visualizations (Bar Chart and Boxplot) to represent the rating analysis, showing average ratings per category and variation in ratings.

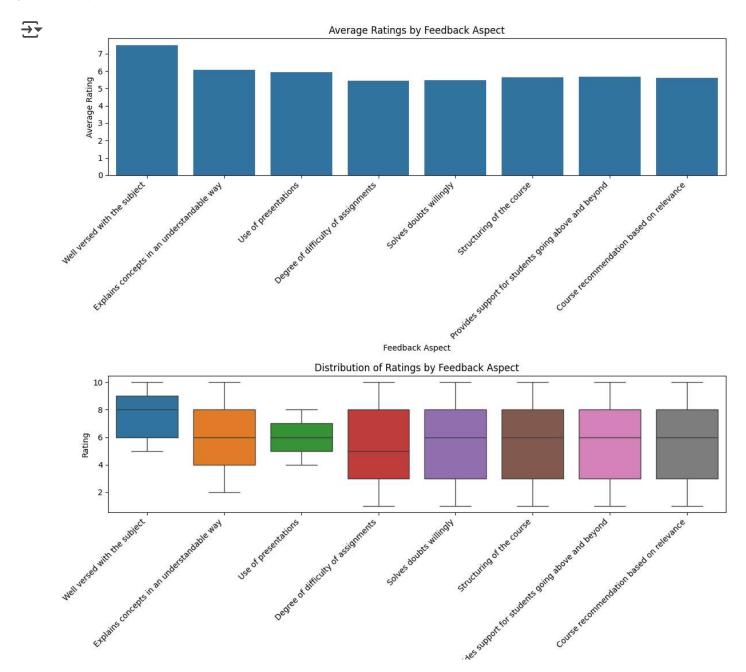
Reasoning: Create a bar chart of average ratings and a boxplot of rating distributions to visualize the analysis results.

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

plt.figure(figsize=(12, 6))
sns.barplot(x=average_ratings.index, y=average_ratings.values)
plt.title('Average Ratings by Feedback Aspect')
plt.xlabel('Feedback Aspect')
```

```
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()

plt.figure(figsize=(12, 6))
sns.boxplot(data=feedback_ratings)
plt.title('Distribution of Ratings by Feedback Aspect')
plt.xlabel('Feedback Aspect')
plt.ylabel('Rating')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```



Perform sentiment analysis

Subtask:

Use NLP tools like TextBlob to analyze the sentiment of the text-based feedback.

Visualize sentiment analysis

Subtask:

Create visualizations (e.g., a Pie Chart) to represent the sentiment distribution of the feedback.

Analyze correlation

Subtask:

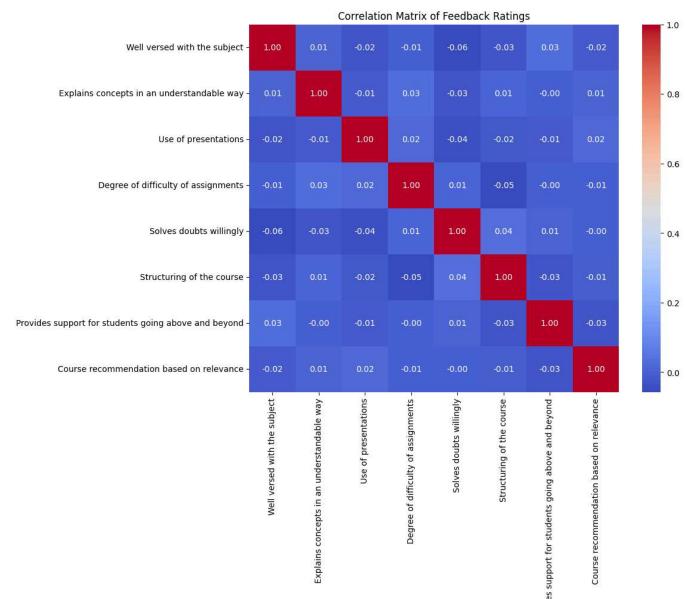
Create a Heatmap to visualize the correlation between different feedback aspects, including ratings and potentially sentiment.

Reasoning: Calculate the correlation matrix and create a heatmap to visualize the correlations between the feedback ratings.

```
correlation_matrix = feedback_ratings.corr()

plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title('Correlation Matrix of Feedback Ratings')
plt.show()
```





Pie chart for course recommendation

Subtask:

Create a pie chart specifically for the average score of "Course recommendation".

Reasoning: Extract the average rating for "Course recommendation based on relevance" and calculate the remaining percentage to create the data for the pie chart. Then, create the pie chart with appropriate labels and title.

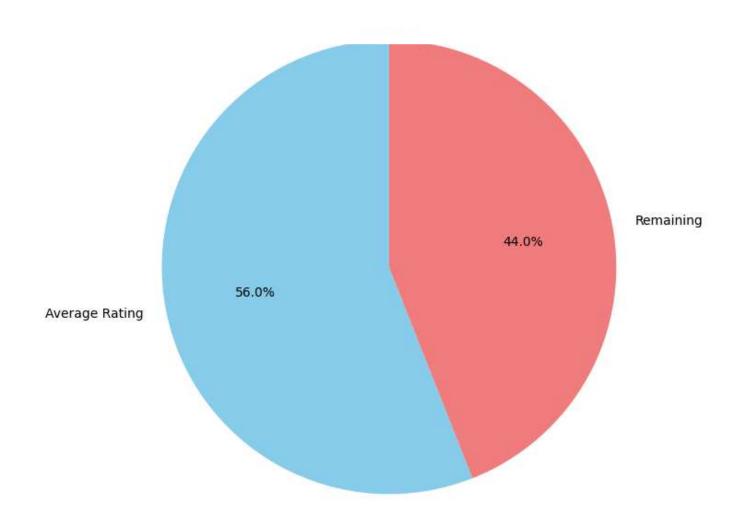
```
course_recommendation_avg = average_ratings["Course recommendation based on relevance"]
remaining_percentage = 10 - course_recommendation_avg # Assuming the rating is out of 10
```

```
sizes = [course_recommendation_avg, remaining_percentage]
labels = ["Average Rating", "Remaining"]
```

```
plt.figure(figsize=(8, 8))
plt.pie(sizes, labels=labels, autopct='%1.1f%%', startangle=90, colors=['skyblue', 'lightcor
plt.title('Average Course Recommendation Score')
plt.show()
```



Average Course Recommendation Score



Summarize and suggest improvements

Subtask:

Summarize the findings from the rating analysis, sentiment analysis, and correlation analysis. Based on these findings, suggest actionable improvements for future events.

Reasoning: Synthesize the findings from the rating analysis and correlation analysis, acknowledging the lack of sentiment analysis, and suggest actionable improvements based on these findings.

```
print("Summary of Findings:")
print("----")
print("Overall student satisfaction based on average ratings appears moderate. The highest-r
print(average ratings.nlargest(3))
print("\nThe lowest-rated aspects (areas for improvement) are:")
print(average_ratings.nsmallest(3))
print("\nSentiment analysis could not be performed as the dataset contained only numerical r
print("\nCorrelation Analysis:")
print("Based on the correlation matrix, there are no strong positive or negative correlation
print("The correlation coefficients are generally low, suggesting that changes in one feedba
print("\nActionable Improvements:")
print("----")
print("Based on the areas for improvement identified:")
print("- Focus on improving the 'Degree of difficulty of assignments'. This could involve re
print("- Address the lower rating for 'Solves doubts willingly'. This might involve encourage
print("- Investigate the reasons behind the lower 'Course recommendation based on relevance'
print("\nGiven the low correlations, improvements in one area are not strongly linked to cha
```

→ Summary of Findings:

Overall student satisfaction based on average ratings appears moderate. The highest-rate Well versed with the subject 7.497502

Explains concepts in an understandable way

6.081918
Use of presentations

5.942058

dtype: float64

The lowest-rated aspects (areas for improvement) are:
Degree of difficulty of assignments 5.430569
Solves doubts willingly 5.474525
Course recommendation based on relevance 5.598402

dtype: float64

Sentiment analysis could not be performed as the dataset contained only numerical rating

Correlation Analysis:

Based on the correlation matrix, there are no strong positive or negative correlations to The correlation coefficients are generally low, suggesting that changes in one feedback

Actionable Improvements:

Based on the areas for improvement identified:

- Focus on improving the 'Degree of difficulty of assignments'. This could involve revie
- Address the lower rating for 'Solves doubts willingly'. This might involve encouraging
- Investigate the reasons behind the lower 'Course recommendation based on relevance' ra

Given the low correlations, improvements in one area are not strongly linked to changes

Summary:

Data Analysis Key Findings

- The average rating for feedback aspects ranged from approximately 5.43 ("Degree of difficulty of assignments") to 7.50 ("Well versed with the subject").
- The top 3 highest-rated aspects (strengths) were "Well versed with the subject" (7.497502), "Explains concepts in an understandable way" (6.081918), and "Use of presentations" (5.942058).
- The bottom 3 lowest-rated aspects (areas for improvement) were "Degree of difficulty of assignments" (5.430569), "Solves doubts willingly" (5.474525), and "Course