

Task

Analyze student event feedback from ["/content/Student_Satisfaction_Survey.csv"](#) to uncover satisfaction trends and suggest improvements using survey data.

Load data

```
import pandas as pd
df = pd.read_csv('/content/Student_Satisfaction_Survey.csv', encoding='latin1')
display(df.head())
display(df.info())
```



	SN	Total Feedback Given	Total Configured	Questions	Weightage 1	Weightage 2	Weightage 3	Weightage 4	Weightage 5	Average/ Percentage	Course Name	Basic C
0	1	1	12	How much of the syllabus was covered in the cl...	0	0	1	0	0	3.00 / 60.00	FY B.VOC FOOD TECHNOLOGY	B.VOC F TECHNOI
1	2	1	12	How well did the teachers prepare for the clas...	0	0	0	0	1	5.00 / 100.00	FY B.VOC FOOD TECHNOLOGY	B.VOC F TECHNOI
2	3	1	12	How well were the teachers able to communicate?	0	0	0	0	1	5.00 / 100.00	FY B.VOC FOOD TECHNOLOGY	B.VOC F TECHNOI
3	4	1	12	The teacher's approach to teaching can best be...	0	0	1	0	0	3.00 / 60.00	FY B.VOC FOOD TECHNOLOGY	B.VOC F TECHNOI
4	5	1	12	Fairness of the internal evaluation process by...	0	0	0	1	0	4.00 / 80.00	FY B.VOC FOOD TECHNOLOGY	B.VOC F TECHNOI

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 580 entries, 0 to 579
Data columns (total 12 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   SN                                    580 non-null    int64
1   Total Feedback Given                 580 non-null    int64
2   Total Configured                    580 non-null    int64
3   Questions                           580 non-null    object
4   Weightage 1                         580 non-null    int64
5   Weightage 2                         580 non-null    int64
6   Weightage 3                         580 non-null    int64
7   Weightage 4                         580 non-null    int64
8   Weightage 5                         580 non-null    int64
9   Average/ Percentage                 580 non-null    object
10  Course Name                         580 non-null    object
11  Basic Course                        580 non-null    object
dtypes: int64(8), object(4)
```

Data cleaning

```
print("Missing values before cleaning:")
print(df.isnull().sum())
print("\nData types before cleaning:")
print(df.dtypes)

df['Average/ Percentage'] = df['Average/ Percentage'].astype(str).str.split('/').str[0].str.split('%').str[0]
df['Average/ Percentage'] = pd.to_numeric(df['Average/ Percentage'], errors='coerce')

df_cleaned = df.drop(columns=['SN', 'Total Feedback Given', 'Total Configured'])

display(df_cleaned.head())
display(df_cleaned.info())
```

```

Missing values before cleaning:
SN                0
Total Feedback Given  0
Total Configured    0
Questions          0
Weightage 1        0
Weightage 2        0
Weightage 3        0
Weightage 4        0
Weightage 5        0
Average/ Percentage 0
Course Name        0
Basic Course       0
dtype: int64

```

Data types before cleaning:

```

SN                int64
Total Feedback Given  int64
Total Configured    int64
Questions          object
Weightage 1        int64
Weightage 2        int64
Weightage 3        int64
Weightage 4        int64
Weightage 5        int64
Average/ Percentage object
Course Name        object
Basic Course       object
dtype: object

```

	Questions	Weightage 1	Weightage 2	Weightage 3	Weightage 4	Weightage 5	Average/ Percentage	Course Name	Basic Course
0	How much of the syllabus was covered in the cl...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY
1	How well did the teachers prepare for the clas...	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY
2	How well were the teachers able to communicate?	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY
3	The teacher's approach to teaching can best be...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY
4	Fairness of the internal evaluation process by...	0	0	0	1	0	4.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 580 entries, 0 to 579
Data columns (total 9 columns):

```

```

#   Column                Non-Null Count  Dtype
---  -
0   Questions            580 non-null    object
1   Weightage 1          580 non-null    int64
2   Weightage 2          580 non-null    int64
3   Weightage 3          580 non-null    int64
4   Weightage 4          580 non-null    int64
5   Weightage 5          580 non-null    int64

```

Sentiment analysis

```

import nltk
from nltk.sentiment.vader import SentimentIntensityAnalyzer

nltk.download('vader_lexicon')

analyzer = SentimentIntensityAnalyzer()

df_cleaned['sentiment_score'] = df_cleaned['Questions'].apply(lambda x: analyzer.polarity_scores(x)['compound'])

display(df_cleaned.head())

```

```
[nltk_data] Downloading package vader_lexicon to /root/nltk_data...
[nltk_data] Package vader_lexicon is already up-to-date!
```

	Questions	Weightage 1	Weightage 2	Weightage 3	Weightage 4	Weightage 5	Average/ Percentage	Course Name	Basic Course	sentiment_score
0	How much of the syllabus was covered in the cl...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.0000
1	How well did the teachers prepare for the clas...	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.2732
2	How well were the teachers able to communicate?	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.2732
3	The teacher's approach to teaching can best be...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.6369

```
from nltk.sentiment.vader import SentimentIntensityAnalyzer
import nltk
```

```
try:
    nltk.data.find('sentiment/vader_lexicon.zip')
except LookupError:
    nltk.download('vader_lexicon')
```

```
analyzer = SentimentIntensityAnalyzer()
```

```
df_cleaned['sentiment_score'] = df_cleaned['Questions'].apply(lambda x: analyzer.polarity_scores(x)['compound'])
```

```
display(df_cleaned.head())
```

```
[nltk_data] Downloading package vader_lexicon to /root/nltk_data...
[nltk_data] Package vader_lexicon is already up-to-date!
```

	Questions	Weightage 1	Weightage 2	Weightage 3	Weightage 4	Weightage 5	Average/ Percentage	Course Name	Basic Course	sentiment_score
0	How much of the syllabus was covered in the cl...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.0000
1	How well did the teachers prepare for the clas...	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.2732
2	How well were the teachers able to communicate?	0	0	0	0	1	5.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.2732
3	The teacher's approach to teaching can best be...	0	0	1	0	0	3.0	FY B.VOC FOOD TECHNOLOGY	B.VOC FOOD TECHNOLOGY	0.6369

Identify satisfaction trends

```
print(df_cleaned.columns)
```

```
Index(['Questions', 'Weightage 1', 'Weightage 2', 'Weightage 3', 'Weightage 4',
      'Weightage 5', 'Average/ Percentage', 'Course Name ', 'Basic Course',
      'sentiment_score'],
      dtype='object')
```

```
df_cleaned.rename(columns={'Course Name ': 'Course Name'}, inplace=True)
```

```
overall_average_sentiment = df_cleaned['sentiment_score'].mean()
print(f"Overall Average Sentiment Score: {overall_average_sentiment:.4f}")
```

```
course_sentiment = df_cleaned.groupby('Course Name')['sentiment_score'].mean().sort_values(ascending=False)
print("\nAverage Sentiment Score by Course Name:")
display(course_sentiment)
```

```
basic_course_sentiment = df_cleaned.groupby('Basic Course')['sentiment_score'].mean().sort_values(ascending=False)
print("\nAverage Sentiment Score by Basic Course:")
display(basic_course_sentiment)

print("\nSentiment Score Distribution (Descriptive Statistics):")
display(df_cleaned['sentiment_score'].describe())

plt.figure(figsize=(10, 6))
sns.histplot(df_cleaned['sentiment_score'], kde=True, bins=20)
plt.title('Distribution of Sentiment Scores')
plt.xlabel('Sentiment Score')
plt.ylabel('Frequency')
plt.grid(True)
plt.show()

question_sentiment = df_cleaned.groupby('Questions')['sentiment_score'].mean().sort_values()

print("\nQuestions with Lowest Average Sentiment Scores:")
display(question_sentiment.head())

print("\nQuestions with Highest Average Sentiment Scores:")
display(question_sentiment.tail())
```

🔗 Overall Average Sentiment Score: 0.2873

Average Sentiment Score by Course Name:

	sentiment_score
Course Name	
FY B.VOC FOOD TECHNOLOGY	0.287335
FY BCOM (ACCOUNTING & FINANCE)	0.287335
FY BCOM (BANKING & INSURANCE)	0.287335
FYBA	0.287335
FYBMS	0.287335
FYBSC	0.287335
M.SC PART - 1 COMPUTER SCIENCE	0.287335
M.SC PART - 2 COMPUTER SCIENCE	0.287335
MA PSYCHOLOGY - 1	0.287335
MA PSYCHOLOGY - 3	0.287335
MSC ANALYTICAL CHEMISTRY SEM I	0.287335
MSC ANALYTICAL CHEMISTRY SEM III	0.287335
MSC DATA SCIENCE - 1	0.287335
MSC DATA SCIENCE - 3	0.287335
MSC INFORMATION TECHNOLOGY - 1	0.287335
MSC INFORMATION TECHNOLOGY - 3	0.287335
MSC MICROBIOLOGY - 1	0.287335
MSC MICROBIOLOGY - 3	0.287335
MSC ORGANIC CHEMISTRY - 3	0.287335
MSC PHYSICS - 3	0.287335
S.Y.B.A.F	0.287335
SY COMPUTER SCIENCE	0.287335
SYBCOM	0.287335
SYBMS	0.287335
SYBSC	0.287335
TYBA	0.287335
TYBCOM	0.287335
TYBMS	0.287335
TYBSC	0.287335

▼ Extract key themes and suggestions
dtype: float64

```
import re
from nltk.corpus import stopwords
```

```
try:
    nltk.data.find('tokenizers/punkt')
except LookupError:
    nltk.download('punkt')

try:
    nltk.data.find('corpora/stopwords')
except LookupError:
    nltk.download('stopwords')

try:
    nltk.data.find('tokenizers/punkt_tab')
except LookupError:
    nltk.download('punkt_tab')
```

```
def preprocess_text(text):
    text = text.lower()
    text = re.sub(r'[^\\w\\s]', '', text)
    tokens = word_tokenize(text)
    tokens = [word for word in tokens if word not in stop_words]
    return ' '.join(tokens)
```

```
display(df_cleaned[['Questions', 'cleaned_questions']].head())
```

	sentiment_score	Questions	cleaned_questions
0		How much of the syllabus was covered in the class?	much syllabus covered class
1	mean 0.287333	How well did the teachers prepare for the class?	well teachers prepare classes
2		How well were the teachers able to communicate?	well teachers able communicate
3	min 0.000000	The teacher's approach to teaching can best be described as...	teachers approach teaching best described
4		Fairness of the internal evaluation process by teachers	fairness internal evaluation process teachers
50%	0.272200		

```
print("Shape of TF-IDF matrix:", tfidf_matrix.shape)
```

Distribution of Sentiment Scores

```
num_clusters = 5
kmeans = KMeans(n_clusters=num_clusters, random_state=42, n_init=10)
clusters = kmeans.fit_predict(tfidf_matrix)
```

```
print(f"\nTop terms per cluster (K={num_clusters}):")
order_centroids = kmeans.cluster_centers_.argsort()[:, ::-1]
terms = tfidf_vectorizer.get_feature_names_out()
```

```
print(" Sample Questions:")
sample_questions = df_cleaned[df_cleaned['cluster_label'] == i]['Questions'].sample(min(3, (df_cleaned['cluster_label'] == i).sum()))
for q in sample_questions:
    print(f"      - {q}")
print("-" * 30)
```

Questions with Lowest Average Sentiment Scores:

Questions

	Fairness of the internal evaluation process by the teachers.	0.0
Top terms per cluster (K=5):	How much of the syllabus was covered in the class?	0.0
Cluster 0:		
Top terms:	Teachers inform you about your expected competencies, course outcomes and program outcomes	0.0
Sample Questions:		
	The institute/ teachers use student-centric methods, such as experiential learning, participative learning and problem-solving methodologies for enhancing learning experiences.	0.0
	- How much of the syllabus was covered in the class?	
	- Your mentor does a necessary follow-up with an assigned task to you	0.0
	The teachers illustrate the concepts through examples and applications.	0.0
Cluster 1:		
Top terms:	learning, skills, opportunities, institute, learn, provides, multiple, grow, institution, internships	
Sample Questions:		
Questions with Highest Average Sentiment Score:	Teachers to inculcate soft skills, life skills and employability skills to make you ready	
	- The institute takes an active interest in promoting internships, student exchange, field visit opportunities for students	
	- The institute/ teachers use student-centric methods, such as experiential learning, participative learning and problem-solving	
Cluster 2:		
Top terms:	Teaching, best, approach, teaching, learning, facilitates, growth	0.5106
Sample Questions:		
	The teaching and mentoring process in your institute can best be described as	0.6369
	- The teaching and mentoring process in your institute can best be described as	
	The institution makes effort to engage students in the monitoring, review and continuous quality improvement of the teaching-learning process.	0.6597
	- The teachers approach to teaching can best be described as	
	- What percentage of teachers use ICT tools such as LCD projectors, Multimedia, etc. while teaching?	
Cluster 3:		
Top terms:	The institute takes an active interest in promoting internships, student exchange, field visit opportunities for students	0.7430
Sample Questions:		
	Teachers encourage you to participate in extracurricular activities.	0.8689
	- How well did the teachers prepare for the classes?	
	Teachers inform you about your expected competencies, course outcomes and program outcomes.	
Cluster 4:		
Top terms:	process, quality, teachinglearning, fairness, evaluation, internal, overall, good, institute, makes	
Sample Questions:		
	- The institution makes effort to engage students in the monitoring, review and continuous quality improvement of the teaching	
	- Fairness of the internal evaluation process by the teachers.	
	- The overall quality of the teaching-learning process in your institute is very good.	
Average Sentiment Score by Cluster:		
	sentiment_score	
cluster_label		
1	0.402975	
4	0.384133	
2	0.376933	
3	0.211400	
0	0.158920	
dtype: float64		

Visualize findings

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

plt.figure(figsize=(10, 6))
sns.barplot(x=cluster_sentiment.index, y=cluster_sentiment.values, palette='viridis')
plt.title('Average Sentiment Score by Cluster')
plt.xlabel('Cluster Label')
plt.ylabel('Average Sentiment Score')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.show()

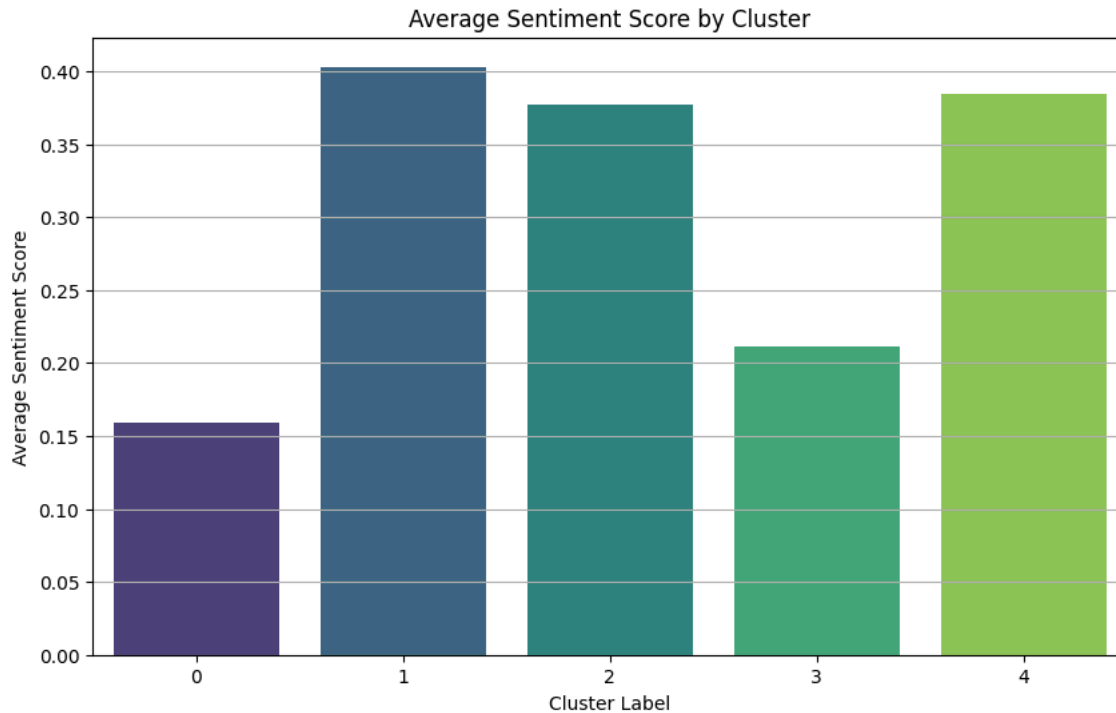
top_10_questions = question_sentiment.tail(10)
bottom_10_questions = question_sentiment.head(10)
combined_questions = pd.concat([bottom_10_questions, top_10_questions])

plt.figure(figsize=(12, 8))
sns.barplot(x=combined_questions.values, y=combined_questions.index, palette='coolwarm')
plt.title('Average Sentiment Score for Highest and Lowest Rated Questions')
plt.xlabel('Average Sentiment Score')
plt.ylabel('Question')
plt.grid(axis='x')
plt.show()
```

```
/tmp/ipython-input-42-2510660229.py:5: FutureWarning:
```

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

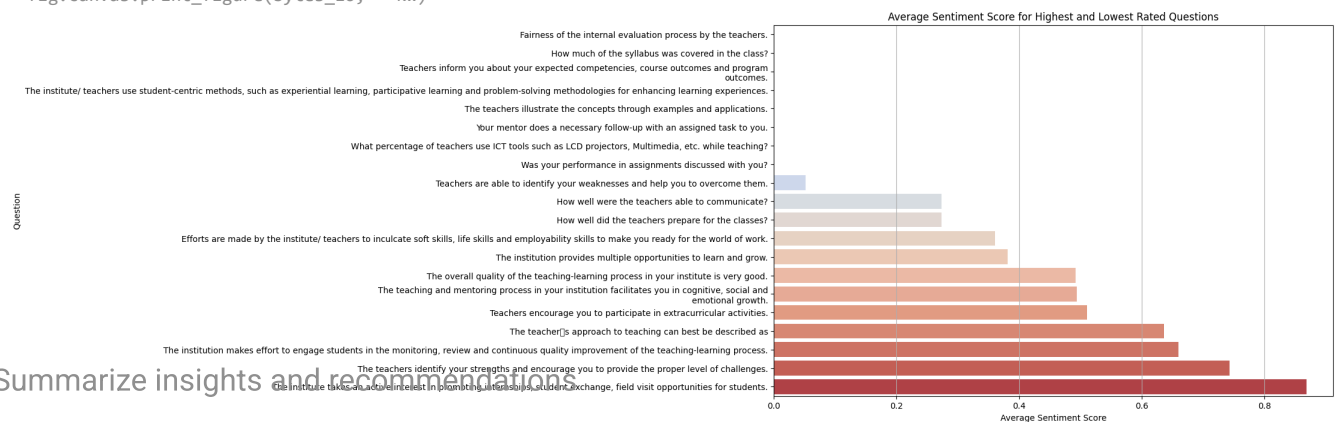
```
sns.barplot(x=cluster_sentiment.index, y=cluster_sentiment.values, palette='viridis')
```



```
/tmp/ipython-input-42-2510660229.py:18: FutureWarning:
```

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

```
sns.barplot(x=combined_questions.values, y=combined_questions.index, palette='coolwarm')  
/usr/local/lib/python3.11/dist-packages/IPython/core/pylabtools.py:151: UserWarning: Glyph 146 (\x92) missing from font(s) DejaVu  
fig.canvas.print_figure(bytes_io, **kw)
```



Summarize insights and recommendations

```
print("--- Overall Sentiment ---")  
print(f"The overall average sentiment score is: {df_cleaned['sentiment_score'].mean():.4f}")  
print("This indicates a generally positive, but not overwhelmingly enthusiastic, sentiment among students.")
```

```
print("\n--- Sentiment by Course and Basic Course ---")  
print("Average sentiment scores are relatively consistent across different Course Names and Basic Courses.")  
print("This suggests that satisfaction levels, based on these survey questions, do not significantly vary by academic program.")
```

```
print("\n--- Sentiment Distribution ---")  
print("The distribution of sentiment scores shows a range from negative/neutral (0.0) to highly positive (0.8689).")  
print("A notable number of responses are neutral or slightly positive, with fewer strongly positive responses.")
```

```
print("\n--- Sentiment by Cluster and Thematic Areas ---")  
print("Analyzing sentiment by cluster, related to the identified themes:")  
display(cluster_sentiment)  
print("\nCluster 1 (highest sentiment) relates to the institute's efforts in providing learning opportunities, skill development, inte  
print("Cluster 4 and 2 also show relatively high sentiment, relating to overall quality, evaluation fairness, teacher approach, and me  
print("Cluster 3 and 0 show lower sentiment, related to teacher communication, preparation, informing about outcomes, teacher support,
```

```
print("\n--- Sentiment for Highest and Lowest Rated Questions ---")  
print("Questions with the highest sentiment highlight areas of strength:")
```

```
display(top_10_questions)
print("\nThese include the institute's active interest in internships, teachers identifying strengths, engaging students in quality im
print("\nQuestions with the lowest sentiment highlight areas needing improvement:")
display(bottom_10_questions)
print("\nThese include fairness of internal evaluation, syllabus coverage, teachers informing about outcomes, use of student-centric m

print("\n--- Actionable Insights and Recommendations ---")
print("Based on the analysis, here are some actionable insights and recommendations:")

print("\nAreas of Strength to Leverage:")
print("- Internships and Opportunities: The high sentiment around internships, student exchange, and field visits suggests these are h
print("- Teacher Engagement and Support (Identifying Strengths): Teachers identifying strengths is perceived positively. Recommendatio
print("- Overall Quality and Teacher Approach: The generally positive sentiment towards overall quality and teacher approach/mentoring

print("\nAreas for Improvement:")
print("- Fairness of Internal Evaluation and Syllabus Coverage: These questions received the lowest sentiment. Recommendation: Review
print("- Teacher Communication and Preparation: Lower sentiment regarding communication and preparation for classes. Recommendation: P
print("- Student-Centric Methods and Concept Illustration: Low sentiment suggests these teaching methods may not be consistently appli
print("- Mentor Follow-up and Assignment Discussion: Low sentiment in these areas. Recommendation: Implement guidelines or training fo
print("- Use of ICT Tools: Low sentiment regarding the percentage of teachers using ICT tools. Recommendation: Invest in and promote t
print("- Informing about Outcomes: Low sentiment regarding teachers informing about expected competencies, course outcomes, and progra

print("\nGeneral Recommendations:")
print("- Gather More Detailed Feedback: The current 'Questions' column is limited. Implement mechanisms for collecting more detailed,
print("- Follow-up on Low-Sentiment Areas: Conduct targeted surveys or focus groups in areas identified with low sentiment to understa
```

--- Overall Sentiment ---
The overall average sentiment score is: 0.2873
This indicates a generally positive, but not overwhelmingly enthusiastic, sentiment among students.

--- Sentiment by Course and Basic Course ---
Average sentiment scores are relatively consistent across different Course Names and Basic Courses.
This suggests that satisfaction levels, based on these survey questions, do not significantly vary by academic program.

--- Sentiment Distribution ---
The distribution of sentiment scores shows a range from negative/neutral (0.0) to highly positive (0.8689).
A notable number of responses are neutral or slightly positive, with fewer strongly positive responses.

--- Sentiment by Cluster and Thematic Areas ---
Analyzing sentiment by cluster, related to the identified themes:

sentiment_score	
cluster_label	
1	0.402975
4	0.384133
2	0.376933
3	0.211400
0	0.158920

dtype: float64

Cluster 1 (highest sentiment) relates to the institute's efforts in providing learning opportunities, skill development, internshi
Cluster 4 and 2 also show relatively high sentiment, relating to overall quality, evaluation fairness, teacher approach, and mento
Cluster 3 and 0 show lower sentiment, related to teacher communication, preparation, informing about outcomes, teacher support, sy

--- Sentiment for Highest and Lowest Rated Questions ---
Questions with the highest sentiment highlight areas of strength:

sentiment_score	
Questions	
How well did the teachers prepare for the classes?	0.2732
Efforts are made by the institute/ teachers to inculcate soft skills, life skills and employability skills to make you ready for the world of work.	0.3612
The institution provides multiple opportunities to learn and grow.	0.3818
The overall quality of the teaching-learning process in your institute is very good.	0.4927
The teaching and mentoring process in your institution facilitates you in cognitive, social and\nemotional growth.	0.4939
Teachers encourage you to participate in extracurricular activities.	0.5106
The teacher's approach to teaching can best be described as	0.6369
The institution makes effort to engage students in the monitoring, review and continuous quality improvement of the teaching-learning process.	0.6597