Importing Relevant Libraries

```
In [63]: import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import gensim
          from gensim import corpora
         from gensim.models import LdaModel
          import pprint
          from textblob import TextBlob
          import nltk
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         from nltk.stem import WordNetLemmatizer
In [64]: # !pip install openpyxl
          # !pip install textblob
         # !pip install gensim
         # nltk.download('punkt')
         # nltk.download('stopwords')
         # nltk.download('wordnet')
         [nltk_data] Downloading package punkt to
         [nltk_data]
                       C:\Users\jagme\AppData\Roaming\nltk_data...
         [nltk_data] Package punkt is already up-to-date!
         [nltk_data] Downloading package stopwords to
                      C:\Users\jagme\AppData\Roaming\nltk_data...
         [nltk_data]
         [nltk_data] Package stopwords is already up-to-date!
         [nltk_data] Downloading package wordnet to
         [nltk_data] C:\Users\jagme\AppData\Roaming\nltk_data...
```

Data Preprocessing

True

Out[64]:

```
In [65]: # Load the Excel file
         excel_file = "AI_Engineer_Dataset_Task_1.xlsx"
         df = pd.read_excel(excel_file)
         # Assuming the text data is in a column named "text_column"
         text_data = df["ParticipantResponse"]
         # Function to preprocess text using NLTK
         def preprocess_text_nltk(text):
             if isinstance(text, str):
                 # Convert text to Lowercase
                 text = text.lower()
                 # Tokenize text using NLTK
                 tokens = word_tokenize(text)
                  # Remove punctuation, special characters, and digits
                 tokens = [word for word in tokens if word.isalpha()]
                  # Remove stopwords using NLTK
                 stop_words = set(stopwords.words('english'))
                  tokens = [word for word in tokens if word not in stop_words]
                 # Lemmatize words using NLTK
                 lemmatizer = WordNetLemmatizer()
                 tokens = [lemmatizer.lemmatize(word) for word in tokens]
                 # Reconstruct the text
                 text = ' '.join(tokens)
             else:
                 # If it's not a string, convert it to an empty string
             return text
         # Apply the preprocessing function to the text data
         text_data_preprocessed_nltk = text_data.apply(preprocess_text_nltk)
         # Display the preprocessed text data
         print(text_data_preprocessed_nltk)
                            disagree
                   strongly disagree
                   strongly disagree
                   strongly disagree
                   strongly disagree
         180964
         180965
                               agree
         180966
                      strongly agree
```

Sentiment Analysis

strongly agree

strongly agree

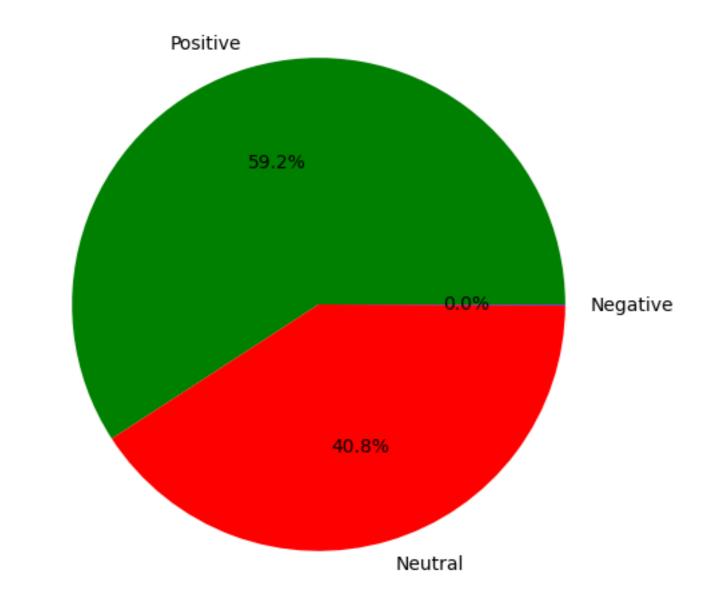
Name: ParticipantResponse, Length: 180969, dtype: object

180967

180968

```
In [66]: # function to assign sentiment
          def get_sentiment_polarity(text):
              analysis = TextBlob(text)
             if analysis.sentiment.polarity > 0:
                  return "Positive"
              elif analysis.sentiment.polarity < 0:</pre>
                  return "Negative"
              else:
                  return "Neutral"
          # Apply sentiment analysis to each comment
         text_data_preprocessed_nltk = text_data_preprocessed_nltk.astype(str) # Ensure all values are strings
         text_data['Sentiment'] = text_data_preprocessed_nltk.apply(get_sentiment_polarity)
          # Calculate the distribution of sentiment
          sentiment distribution = text data['Sentiment'].value counts()
          # Create a pie chart to visualize sentiment distribution
          plt.figure(figsize=(8, 6))
          sentiment_distribution.plot(kind='pie', autopct='%1.1f%%', colors=['green', 'red', 'blue'])
          plt.title('Sentiment Distribution of Feedback Comments')
          plt.ylabel('')
          plt.show()
         # Display summary statistics
          print("Sentiment Distribution Summary:")
          print(sentiment_distribution)
         C:\Users\jagme\AppData\Local\Temp\ipykernel_1576\2579182003.py:41: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
```

text_data['Sentiment'] = text_data_preprocessed_nltk.apply(get_sentiment_polarity) Sentiment Distribution of Feedback Comments



Sentiment Distribution Summary: ParticipantResponse Positive 107065 Neutral 73819 Negative Name: count, dtype: int64

Topic Modeling

print("Identified Topics:")

for topic in topics: print(topic)

Identified Topics:

```
In [67]: # Tokenize the preprocessed text data
         tokenized_data = text_data_preprocessed_nltk.apply(lambda x: x.split())
         # Create a dictionary from the tokenized data
         dictionary = corpora.Dictionary(tokenized_data)
         # Create a document-term matrix (corpus)
         corpus = [dictionary.doc2bow(text) for text in tokenized_data]
         # Define the number of topics for LDA
         num_topics = 5
In [68]: # Create the LDA model
         lda_model = LdaModel(corpus, num_topics=num_topics, id2word=dictionary, passes=5, random_state=42)
         # Print the topics and their representative keywords
         topics = lda_model.print_topics(num_words=5)
```

```
Insights and Recommendations
In [69]: print("\nTopic Modeling Insights:")
        topics = lda_model.print_topics(num_words=5)
        for i, topic in enumerate(topics):
           print(f"Topic {i+1}: {topic}")
```

```
Topic Modeling Insights:
Topic 1: (0, '0.062*"course" + 0.040*"good" + 0.022*"dr" + 0.019*"thank" + 0.019*"best"')
Topic 2: (1, '0.573*"agree" + 0.422*"strongly" + 0.000*"sss" + 0.000*"لازيوجد"*0.000 + "ن"')
Topic 3: (2, '0.031*"0.024 + "في" + 0.024 + "لا" + 0.027 + "لا" + 0.027 + "لا" + 0.024 + "
Topic 4: (3, '0.567*"disagree" + 0.408*"strongly" + 0.000*"sss" + 0.000*"'لزيوجد"*0.000 + "ن"')
Topic 5: (4, '0.936*"neutral" + 0.004*"none" + 0.000*"sss" + 0.000*"' (پوجد "*0.000 + "ن" )
```

(0, '0.062*"course" + 0.040*"good" + 0.022*"dr" + 0.019*"thank" + 0.019*"best"') (1, '0.573*"agree" + 0.422*"strongly" + 0.000*"sss" + 0.000*"لازبوجد"*0.000 + "ن")

(4, '0.936*"neutral" + 0.004*"none" + 0.000*"sss" + 0.000*"'لزيوجد"*0.000 + "نا")

(3, '0.567*"disagree" + 0.408*"strongly" + 0.000*"sss" + 0.000*" + 0.000 + "الزيوجد" + 0.000 (3, '0.567*"disagree" + 0.408*"strongly" + 0.000*"sss" + 0.000*"

(2, '0.031*"0.024 + "في" + 0.023*"و" + 0.024 + "في" + 0.023*" (2, '0.031*" (2, '0.031*" (2, '0.031*" (2, '0.031*" (2, '0.024 + "في" + 0.023*" (2, '0.024 + "في" (2, '0.031*" (2, '0.024 + "في" + 0.023*" (2, '0.024 + "في" + 0.023*" (2, '0.024 + "في" (2, '0.024 + "b) (2, '0

In [70]: positive_count = sentiment_distribution.get("Positive", 0) negative_count = sentiment_distribution.get("Negative", 0) neutral_count = sentiment_distribution.get("Neutral", 0) In [71]: print("\nRecommendations:")

```
if positive_count > negative_count:
   print("Based on the sentiment analysis, high number of positive comments are identified , indicating areas of strength.")
   print("Based on the sentiment analysis, high number of Negative comments are identified, indicating potential areas for improvement.")
```

Recommendations: Based on the sentiment analysis, there are more positive comments, indicating areas of strength.