

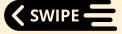


# Sentiment Analysis using Python

for Beginners









In this guide, we will walk through the process of performing sentiment analysis and generating a word cloud using comments from my LinkedIn post.







## **Prerequisites**

- Python installed on your system.
- Jupyter Notebook or any Python IDE for executing the code.
- The comments from your LinkedIn post saved as a CSV.









#### **Load the Comments Data**

```
import pandas as pd

# Load the data
df = pd.read_csv('/kaggle/input/post-comments/linkedin_comments.csv')

#remove commentor name for obvious reason
df = df.drop('commentor name',axis=1)

# Display the first few rows of the data
df.head()
```

```
comment text

O A section on triggers and would be perfect!

1 AJ Mieskolainen; You should take things positi...

2 Amazing! 
3 Cheat sheets are the best! I use them all the ...

4
```







## **Perform Sentiment Analysis**







```
## negative comments
df[df['sentiment']== 'negative'].head(10)

#Textblob is good for simple sentiment analysis,
#however, sometime makes mistakes.
```

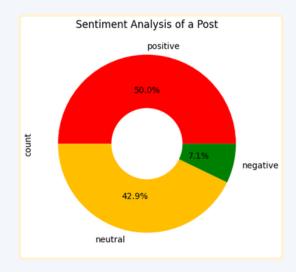
	comment text	polarity	sentiment
7	Gracias por compartir; precisamente estoy estu	-1.00000	negative
25	Well done for covering a broad base and the re	-0.46875	negative







#### Let's show it on bar chart









### **Generate a Word Cloud**

```
from wordcloud import WordCloud import matplotlib.pyplot as plt

# Combine all comments into a single string text = " ".join(comment for comment in df['comment text'].dropna())

# Generate word cloud wordcloud = WordCloud(background_color="white", width=800, height=400).generate(text)

# Display the generated word cloud plt.figure(figsize=(10, 5)) plt.imshow(wordcloud, interpolation='bilinear') plt.axis('off') plt.show()
```









### **Conclusion**

By now, you should have a clear idea about the sentiments of the comments on your LinkedIn post and the most frequently used words in those comments.

This is a basic guide to sentiment analysis and word cloud generation.

For more accurate sentiment analysis, you might consider using more sophisticated models or libraries, training on domain-specific data, or incorporating additional preprocessing steps.







## **Appendix**

#### Using BERT for better sentiment analysis

```
import torch
import pandas as pd
from transformers import BertTokenizer, BertForSequenceClassification
from torch.nn.functional import softmax
model = BertForSequenceClassification.from_pretrained('bert-base-uncased', num_labels=3)
def get_sentiment(text):
    inputs = tokenizer(text, return_tensors="pt", truncation=True,
padding=True, max_length=512)
   with torch.no_grad():
   probs = softmax(outputs[0], dim=1)
    label = torch.argmax(probs, dim=1)
       sentiments = {0: 'Negative', 1: 'Neutral', 2: 'Positive'}
       return sentiments[label.item()]
df['sentiment'] = df['comment text'].apply(lambda x: get_sentiment(str(x)))
df[['comment text', 'sentiment']].head(5)
```







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