pip install nltk pandas matplotlib seaborn

⇒ Show hidden output

```
# import the necessary library
import pandas as pd
import pandas as pd
import nltk
import seaborn as sns
import matplotlib.pyplot as plt
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.stem import WordNetLemmatizer
from collections import Counter

nltk.download('punkt_tab')
nltk.download('stopwords')
nltk.download('wordnet')
```

[nltk_data] Downloading package punkt_tab to /root/nltk_data...
[nltk_data] Unzipping tokenizers/punkt_tab.zip.
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.
[nltk_data] Downloading package wordnet to /root/nltk_data...
True

df = pd.read_csv("/content/customer_support_tickets.csv.zip")

df.head()

₹		Ticket ID	Customer Name	Customer Email	Customer Age	Customer Gender	Product Purchased	Date of Purchase	Ticket Type	Ticket Subject	
	0	1	Marisa Obrien	carrollallison@example.com	32	Other	GoPro Hero			Product setup	I'm {proc
	1	2	Jessica Rios	clarkeashley@example.com	42	Female	LG Smart TV	2021 - 05- 22	Technical issue	Peripheral compatibility	l'm {proc
	2	3	Christopher Robbins	gonzalestracy@example.com	48	Other	Dell XPS	2020 - 07 - 14	Technical issue	Network problem	l'm fa
	3	4	Christina Dillon	bradleyolson@example.org	27	Female	Microsoft Office	2020-11- 13	Billing inquiry	Account access	I'm {proc
	4	5	Alexander Carroll	bradleymark@example.com	67	Female	Autodesk AutoCAD	2020 - 02 - 04	Billing inquiry	Data loss	l'm {proc

Next steps: Generate code with df View recommended plots New interactive sheet

df.info()

```
<<class 'pandas.core.frame.DataFrame'>
   RangeIndex: 8469 entries, 0 to 8468
```

Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype					
0	Ticket ID	8469 non-null	int64					
1	Customer Name	8469 non-null	object					
2	Customer Email	8469 non-null	object					
3	Customer Age	8469 non-null	int64					
4	Customer Gender	8469 non-null	object					
5	Product Purchased	8469 non-null	object					
6	Date of Purchase	8469 non-null	object					
7	Ticket Type	8469 non-null	object					
8	Ticket Subject	8469 non-null	object					
9	Ticket Description	8469 non-null	object					
10	Ticket Status	8469 non-null	object					
11	Resolution	2769 non-null	object					
12	Ticket Priority	8469 non-null	object					
13	Ticket Channel	8469 non-null	object					
14	First Response Time	5650 non-null	object					
15	Time to Resolution	2769 non-null	object					
16	Customer Satisfaction Rating	2769 non-null	float64					
dt (1tc4/1)								

dtypes: float64(1), int64(2), object(14)

memory usage: 1.1+ MB

missing values
df.isnull()



•	Ticket ID	Customer Name	Customer Email	Customer Age	Customer Gender	Product Purchased	Date of Purchase	Ticket Type	Ticket Subject	Ticket Description	Ticket Status	Re
0	False	False	False	False	False	False	False	False	False	False	False	
1	False	False	False	False	False	False	False	False	False	False	False	
2	False	False	False	False	False	False	False	False	False	False	False	
3	False	False	False	False	False	False	False	False	False	False	False	
4	False	False	False	False	False	False	False	False	False	False	False	
8464	False	False	False	False	False	False	False	False	False	False	False	
8465	False	False	False	False	False	False	False	False	False	False	False	
8466	False	False	False	False	False	False	False	False	False	False	False	
8467	False	False	False	False	False	False	False	False	False	False	False	
8468	False	False	False	False	False	False	False	False	False	Fa l se	False	

8469 rows × 17 columns

```
#handling missing values

df = df.assign(
    Resolution=df['Resolution'].fillna("Not Provided"),
    Time_to_Resolution=df['Time to Resolution'].fillna("Not Provided"),
    Customer_Satisfaction_Rating=df['Customer Satisfaction Rating'].fillna(-1),
    First_Response_Time=df['First Response Time'].fillna("Not Provided")
)
# assign() creates a new dataframe and assign the new cleaned column
```

2 3

Low

```
# convert date to datetime data type
df['Date of Purchase'] = pd.to_datetime(df['Date of Purchase'], errors='coerce')
df['First Response Time'] = pd.to_datetime(df['First Response Time'], errors='coerce')
df['Time to Resolution'] = pd.to_datetime(df['Time to Resolution'], format="%Y-%m-%d %H:%M:%S", errors='coerce')
# formatting
df['Customer Gender'] = df['Customer Gender'].str.capitalize()
df['Ticket Description'] = df['Ticket Description'].str.replace(r'{.*?}', 'the product', regex=True)
# removing duplicates
df.drop_duplicates(inplace=True)
# save the cleaned dataset
df.to_csv("cleaned_customer_support_tickets.csv", index=False)
# load the cleaned dataset
df = pd.read_csv("cleaned_customer_support_tickets.csv") # Load the cleaned dataset
print(df.head()) # Preview the data
print(df.info()) # Check for missing values & data types
    3 I'm having an issue with the the product. Plea...
    4 I'm having an issue with the the product. Plea...
                   Ticket Status
                                                                     Resolution \
     0 Pending Customer Response
                                                                   Not Provided
                                                                   Not Provided
    1 Pending Customer Response
     2
                                  Case maybe show recently my computer follow.
    3
                          Closed Try capital clearly never color toward story.
    4
                          Closed
                                                    West decision evidence bit.
      Ticket Priority Ticket Channel First Response Time  Time to Resolution \
             Critical Social media 2023-06-01 12:15:36
                                                                           NaN
                                Chat 2023-06-01 16:45:38
    1
             Critical
```

Low Social media 2023-06-01 11:14:38 2023-06-01 18:05:38

Low Social media 2023-06-01 07:29:40 2023-06-01 01:57:40

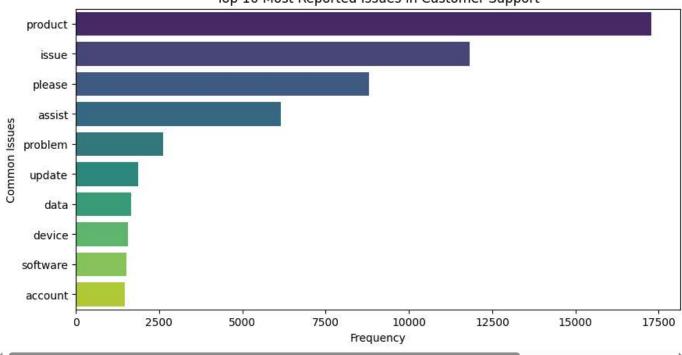
Email 2023-06-01 00:12:42 2023-06-01 19:53:42

```
ортест
         pate of Purchase
                                       8469 non-null
     ь
         Ticket Type
                                       8469 non-null
                                                       object
        Ticket Subject
                                       8469 non-null
                                                       object
        Ticket Description
                                       8469 non-null
                                                       object
                                       8469 non-null
     10 Ticket Status
                                                       object
     11 Resolution
                                       8469 non-null
                                                       object
                                       8469 non-null
     12 Ticket Priority
                                                       object
     13 Ticket Channel
                                       8469 non-null
                                                       object
     14 First Response Time
                                       5650 non-null
                                                       object
                                       2769 non-null
     15 Time to Resolution
                                                       object
         Customer Satisfaction Rating 2769 non-null
                                                       float64
     17 Time_to_Resolution
                                       8469 non-null
                                                        object
     18 Customer_Satisfaction_Rating 8469 non-null
                                                       float64
     19 First_Response_Time
                                       8469 non-null
                                                       object
     dtypes: float64(2), int64(2), object(16)
     memory usage: 1.3+ MB
     None
# preprocess the data
def clean_text(text):
    if pd.isna(text): # Handle missing values
        return ""
   text = text.lower() # Convert to lowercase
   words = word tokenize(text) # Tokenization
   words = [word for word in words if word.isalpha()] # Remove punctuation & numbers
   words = [word for word in words if word not in stopwords.words('english')] # Remove stopwords
   lemmatizer = WordNetLemmatizer()
   words = [lemmatizer.lemmatize(word) for word in words] # Lemmatization
   return " ".join(words)
df["Cleaned_Issue"] = df["Ticket Description"].apply(clean_text)
# finding most common issue
all words = " ".join(df["Cleaned_Issue"]).split() # Combine all text
common_issues = Counter(all_words).most_common(10) # Find top 10 most common words
print(common issues)
    [('product', 17284), ('issue', 11829), ('please', 8810), ('assist', 6147), ('problem', 2609), ('update', 1865),
# frequent reported issues
import matplotlib.pyplot as plt
import seaborn as sns
from collections import Counter
# Count the most common words
all words = " ".join(df["Cleaned Issue"]).split()
common_issues = Counter(all_words).most_common(10) # Top 10
# Convert to DataFrame for visualization
common df = pd.DataFrame(common issues, columns=['Issue', 'Count'])
# Plot
plt.figure(figsize=(10,5))
sns.barplot(data=common_df, x="Count", y="Issue",hue="Issue", palette="viridis")
plt.xlabel("Frequency")
plt.ylabel("Common Issues")
plt.title("Top 10 Most Reported Issues in Customer Support")
plt.show()
```

3/30/25, 10:00 PM Second Task - Colab

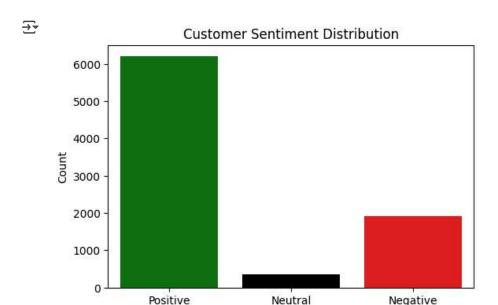


Top 10 Most Reported Issues in Customer Support



```
# perform a sentiment analysis
from nltk.sentiment import SentimentIntensityAnalyzer
nltk.download("vader_lexicon")
sia = SentimentIntensityAnalyzer()
→ [nltk_data] Downloading package vader_lexicon to /root/nltk_data...
# sentiment analysis on customer feedback
def get_sentiment(text):
    if pd.isna(text) or text.strip() == "":
        return "Neutral" # Handle empty text
    score = sia.polarity_scores(text)["compound"]
    if score >= 0.05:
       return "Positive"
    elif score <= -0.05:
        return "Negative"
    else:
        return "Neutral"
df["Sentiment"] = df["Ticket Description"].apply(get_sentiment)
# sentiment visuals
plt.figure(figsize=(6,4))
sns.countplot(data=df, x="Sentiment", hue="Sentiment", palette=["green", "black", "red"])
plt.xlabel("Sentiment Category")
plt.ylabel("Count")
plt.title("Customer Sentiment Distribution")
plt.show()
# Calculate sentiment counts
sentiment_counts = df["Sentiment"].value_counts()
# Print percentage breakdown
sentiment_percentages = (sentiment_counts / sentiment_counts.sum()) * 100
```

print(sentiment_percentages)



Sentiment
Positive 73.149132
Negative 22.659110
Neutral 4.191758
Name: count. dtvpe: float64

Convert 'Cleaned_Issue' to lowercase for consistency

df["Cleaned_Issue"] = df["Cleaned_Issue"].str.lower()

Define top issues as lowercase

top_issues = ["product", "issue", "please", "assist", "problem", "update", "data", "device", "software", "account"]

Filter rows where 'Cleaned_Issue' contains any top issue keyword

filtered_df = df[df["Cleaned_Issue"].apply(lambda x: any(word in x for word in top_issues))]

Sentiment Category

Group and count sentiment distribution for filtered issues
issue_sentiment = filtered_df.groupby(["Cleaned_Issue", "Sentiment"]).size().unstack().fillna(0)

Print result

print(issue sentiment)

$\overline{\Rightarrow}$	Sentiment	Negative	Neutral	\
	Cleaned_Issue			
	accidentally deleted important data product way	1.0	0.0	
	accidentally deleted important data product way	0.0	0.0	
	accidentally deleted important data product way	1.0	0.0	
	accidentally deleted important data product way	1.0	0.0	
	accidentally deleted important data product way	1.0	0.0	
	•••			
	unable access product account keep displaying c	1.0	0.0	
	unable access product account keep displaying c	1.0	0.0	
	unable access product account keep displaying c	4.0	0.0	
	unable access product account keep displaying c	0.0	0.0	
	unable access product account keep displaying c	0.0	1.0	
	Sentiment	Positive		
	Cleaned Issue			
	accidentally deleted important data product way	0.0		
	accidentally deleted important data product way	1.0		
	accidentally deleted important data product way	0.0		
	accidentally deleted important data product way	0.0		
	, , , , , , , , , , , , , , , , , , , ,			
	accidentally deleted important data product way	0.0		

unable access product account keep displaying c... 0.0
unable access product account keep displaying c... 0.0
unable access product account keep displaying c... 0.0
unable access product account keep displaying c... 1.0
unable access product account keep displaying c... 0.0

[7902 rows x 3 columns]