

Data Collection and Preprocessing

Project Title : Amazon Instruments Review

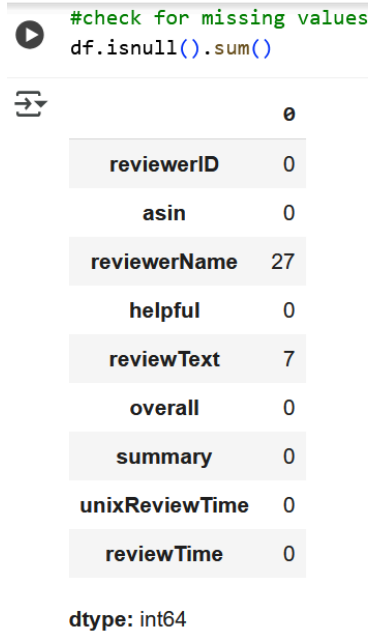

Summarizes the preprocessing steps applied to the Amazon Instrument Reviews dataset to ensure clean and consistent input for machine learning models.

Preprocessing steps :

Section	Description
Data Overview	Loaded CSV dataset, checked missing values, and reviewed class distribution.
Handling Missing Values	Filled missing reviewText with empty strings and reviewerName with " ".
Concatenation	Combined reviewText and summary into a single column reviews.
Labelling	Assigned labels: Positive (rating > 3), Negative (rating < 3), Neutral (rating = 3).
Text Cleaning	Lowercasing, punctuation removal, number removal, URL removal, newline removal.
Text Preprocessing	Tokenized reviews, removed stopwords, and applied WordNet Lemmatizer.
Polarity Scores	Calculated polarity of reviews using TextBOB for strength
Feature Engineering	Generated features: review length, word count, n-grams and dropped columns.
Encoding	Encoded target sentiment labels using LabelEncoder.
Vectorization	Applied TF-IDF with Unigram,Bigram, Trigram (max 5000 features).

Balancing	Used SMOTE to oversample minority classes.
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Data Preprocessing Code Screenshots :

Section	CODE
<p>Data Overview</p> <pre>#check for missing values df.isnull().sum()</pre>  <pre>dtype: int64</pre> <p>W</p>	<pre>import numpy as np import pandas as pd import zipfile import os # Extract both files from the zip with zipfile.ZipFile('archive.zip', 'r') as zip_ref: # List all files in the zip print("Files in zip:") for file_info in zip_ref.filelist: print(f"- {file_info.filename}") # Extract both files zip_ref.extractall('extracted_files') print("Files extracted successfully!")</pre> <p>Files in zip: - Musical_Instruments_5.json - Musical_instruments_reviews.csv Files extracted successfully!</p> <pre>df = pd.read_csv('extracted_files/Musical_instruments_reviews.csv') df.head(10)</pre> 

Handling missing values

```
#fill missing values with ""
df['reviewText'] = df['reviewText'].fillna("")
```

```
df.isnull().sum()
```




	0
reviewerID	0
asin	0
reviewerName	27
helpful	0
reviewText	0
overall	0
summary	0
unixReviewTime	0
reviewTime	0

Concatenation

Concatenate review text and Summary columns

```
df['reviews'] = df['reviewText'] + ' ' + df['summary']
df.drop(columns = ["reviewText", "summary"], axis = 1, inplace = True)
df
```



	reviewerID	asin	reviewerName	helpful	overall	unixReviewTime	reviewTime	reviews
0	A28PQ2OUZRIU	1384719342	cosandra lu "Yeah, well, that's just like, u...	[0, 0]	5.0	1303545600	02/28/2014	Not much to write about here, but it's done exact...
1	A14VAT5EAX3D95	1384719342	Jake	[13, 14]	5.0	1363392000	03/16/2013	The product does exactly as it should and is q...
2	A195EZSQOW3E21	1384719342	Rick Bennette "Rick Bennette"	[1, 1]	5.0	1377648000	08/28/2013	The primary job of this device is to block the...
3	A2CO0NNG1ZQOG2	1384719342	RustyBill "Sunday Rocker"	[0, 0]	5.0	1392336000	02/14/2014	Nice windscreen protects my MXL mic and preven...
4	A94QU4C9OB1AX	1384719342	SEAN MASLANKA	[0, 0]	5.0	1392940800	02/21/2014	This pop filter is great. It looks and perform...
...
10256	A1482YH3Z3XMP	B00JBVX0XC	Lorrie M. Adams	[0, 0]	5.0	1405814400	07/20/2014	Great, just as expected. Thank to all from e...

Labelling

Labeling product based on Ratings

```
def Labelling(Rows) :
    if(Rows["overall"] > 3.0) :
        Label = "Positive"
    elif(Rows["overall"] < 3.0) :
        Label = "Negative"
    else :
        Label = "Neutral"
    return Label
```

```
df["sentiment"] = df.apply(Labelling, axis = 1)
df["sentiment"].value_counts().plot(kind = 'bar', color = "blue")
plt.title("Rating")
plt.xlabel("Sentiments", color = "green", fontsize = 10, labelpad = 15)
plt.xticks(rotation = 0)
plt.ylabel("Amount of sentiments")
plt.show()
```

Text Cleaning

Text cleaning

```
import re
import string

def Text_cleaning(Text):

    Text = Text.lower() # 1: lower case
    punc = str.maketrans('', '', string.punctuation)
    Text = Text.translate(punc) # 2: Remove punctuations

    Text = re.sub(r'\d+', '', Text) # 3: Remove numbers
    Text = re.sub(r'https?://\S+|www\.\S+', '', Text) # 4: Remove links
    Text = re.sub('\n', '', Text) # 5: Delete new lines

    return Text
```

Text Preprocessing

Text Preprocessing

```
import nltk
import nltk.corpus
from nltk.corpus import stopwords # Import stopwords here
nltk.download("punkt")
nltk.download("stopwords")
nltk.download("wordnet")
nltk.download("punkt_tab")
from nltk.stem import WordNetLemmatizer
```

```
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt.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...
[nltk_data] Downloading package punkt_tab to /root/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt_tab.zip.
```

Text Preprocessing function

```
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from nltk.tokenize import word_tokenize

def Text_Preprocessing(Text):
    # Tokenization
    tokens = word_tokenize(Text)

    # Remove stop words
    stop_words = set(stopwords.words('english'))
    tokens = [word for word in tokens if word not in stop_words]

    # Lemmatization
    lemmatizer = WordNetLemmatizer()
    tokens = [lemmatizer.lemmatize(word) for word in tokens]

    # Join tokens back into a string
    return " ".join(tokens)
```

```
df["reviews"] = df["reviews"].apply(lambda Text : Text_cleaning(Text))
df["reviews"] = df["reviews"].apply(lambda Text : Text_Preprocessing(Text))
df.head()
```

Polarity Scores

```
from textblob import TextBlob
```

```
df["Polarity"] = df["reviews"].map(lambda Text : TextBlob(Text).polarity)
df["Polarity"].plot(kind="hist", bins=20, edgecolor="black", linewidth=1, color="orange", figsize=(10,5))

plt.title("Polarity Score in Reviews", fontsize=15, pad=20)
plt.xlabel("Polarity", labelpad=5, color="red")
plt.ylabel("Amount of Reviews", labelpad=20, color="green")
plt.show()
```

Feature Engineering

Review length

```
df["length"] = df["reviews"].apply(lambda x: len(x.split()))

# Use plt.hist directly
plt.figure(figsize=(10,5))
plt.hist(df["length"], bins=20, edgecolor="blue", linewidth=1, color="orange")
plt.title("Length of Reviews", color="blue", pad=20)
plt.xlabel("Length", labelpad=15, color="red")
plt.ylabel("Frequency")
plt.show()
```

Word Counts

```
df["Word_count"] = df["reviews"].apply(lambda x: len(x.split()))
plt.hist(df["Word_count"])
```

```
[array([9.486e+03, 6.418e+02, 1.588e+02, 3.908e+01, 1.000e+01, 7.000e+00,
        4.000e+00, 1.000e+00, 2.000e+00, 1.000e+00]),
 array([ 2. , 110.4, 218.8, 327.2, 435.6, 544. , 652.4, 760.8,
        869.2, 977.6, 1086. ]),
 <BarContainer object of 10 artists>)
```

	<pre># N-Gram Analysis def GramAnalysis(Corpus, Gram, N): Vectorizer = CountVectorizer(stop_words="english", ngram_range=(Gram, Gram)) ngram_matrix = Vectorizer.fit_transform(Corpus) # N-Gram Frequency Counts = ngram_matrix.sum(axis=0) # List of words words = [(word, Counts[0, idx]) for word, idx in Vectorizer.vocabulary_.items()] # Sort Descending words = sorted(words, key=lambda x: x[1], reverse=True) return words[:N] # Filter the platforms Based on Sentiments Positive = df[df["sentiment"]=="Positive"].dropna() Negative = df[df["sentiment"]=="Negative"].dropna() Neutral = df[df["sentiment"]=="Neutral"].dropna()</pre> <pre># Feature Engineering columns_to_keep = ['reviews', 'sentiment'] # Add other columns you've created df = df[columns_to_keep] df.head()</pre> <table><tr><th></th><th>reviews</th><th>sentiment</th></tr><tr><td>0</td><td>much write exactly supposed filter pop sound r...</td><td>Positive</td></tr><tr><td>1</td><td>product exactly quite affordable! realized dou...</td><td>Positive</td></tr><tr><td>2</td><td>primary job device block breath would otherwis...</td><td>Positive</td></tr><tr><td>3</td><td>nice windscreen protects mxl mic prevents pop ...</td><td>Positive</td></tr><tr><td>4</td><td>pop filter great look performs like studio fil...</td><td>Positive</td></tr></table>		reviews	sentiment	0	much write exactly supposed filter pop sound r...	Positive	1	product exactly quite affordable! realized dou...	Positive	2	primary job device block breath would otherwis...	Positive	3	nice windscreen protects mxl mic prevents pop ...	Positive	4	pop filter great look performs like studio fil...	Positive
	reviews	sentiment																	
0	much write exactly supposed filter pop sound r...	Positive																	
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3	nice windscreen protects mxl mic prevents pop ...	Positive																	
4	pop filter great look performs like studio fil...	Positive																	
Encoding	<pre># Encoding Our Target Variables import warnings warnings.filterwarnings('ignore') from sklearn.preprocessing import LabelEncoder encoder = LabelEncoder() df['sentiment'] = encoder.fit_transform(df['sentiment']) df['sentiment'].value_counts()</pre> <table><tr><th></th><th>count</th></tr><tr><th>sentiment</th><td></td></tr><tr><td>2</td><td>9022</td></tr><tr><td>1</td><td>772</td></tr><tr><td>0</td><td>467</td></tr></table> <p>dtype: int64</p>		count	sentiment		2	9022	1	772	0	467								
	count																		
sentiment																			
2	9022																		
1	772																		
0	467																		
Vectorization	<pre># TF-IDF Vectorization from sklearn.feature_extraction.text import TfidfVectorizer TF_IDF = TfidfVectorizer(max_features = 5000, ngram_range = (1,3)) X = TF_IDF.fit_transform(df['reviews']).toarray() X.shape Y = df['sentiment'] Counter(Y)</pre> <p>Counter({2: 9022, 1: 772, 0: 467})</p>																		
Balancing	<pre># Resampling our Dataset (to Balance) from imblearn.over_sampling import SMOTE Balancer = SMOTE(random_state=42) X_final, y_final = Balancer.fit_resample(X, Y) Counter(y_final)</pre> <p>Counter({2: 9022, 1: 9022, 0: 9022})</p>																		

