dataanaytisc

June 7, 2024

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0.1 ( )
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[3]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     data_path = './data/reviews_with_sentiment.csv'
     df = pd.read_csv(data_path)
     POS = ['ADJ', 'ADV', 'INTJ', 'PROPN', 'NOUN', 'VERB']
     MAX\_TERMS\_IN\_DOC = 5
     NGRAM = 1
     MAX_DF = 1.0
     MIN_DF = 0.0
     NUM_VOCAB = 10000
     TOP_K = 20
     LAMBDA = 5 #
              Bag-of-Words
     def preprocess_data(df):
         import spacy
         from sklearn.feature_extraction.text import CountVectorizer
         import itertools
         nlp = spacy.load('ja_ginza')
         def flatten(*lists) -> list:
             res = []
             for l in list(itertools.chain.from_iterable(lists)):
                 for e in 1:
                     res.append(e)
             return res
         def remove_duplicates(1):
             d = \{\}
             for e in 1:
```

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d[e[0]] = e[1]
        return list(d.items())
    df["doc"] = [nlp(review) for review in df["review"]]
    bows = \{\}
    cvs = \{\}
    for sentiment in df["sentiment"].unique():
        tokens = \Pi
        for doc in df[df["sentiment"] == sentiment]["doc"]:
            similarities = [(token.similarity(doc), token.lemma) for token in_
 →doc if token.pos_ in POS]
            similarities = remove_duplicates(similarities)
            similarities = sorted(similarities, key=lambda sim: sim[1],__
 →reverse=True)[:MAX_TERMS_IN_DOC]
            tokens.append([similarity[1] for similarity in similarities])
        cv = CountVectorizer(ngram_range=(1, NGRAM), max_df=MAX_DF,__

min_df=MIN_DF, max_features=NUM_VOCAB)
        bows[sentiment] = cv.fit_transform(flatten(tokens)).toarray()
        cvs[sentiment] = cv
    return bows, cvs
bows, cvs = preprocess data(df)
term_frequencies = {}
for sentiment in df["sentiment"].unique():
    bow = bows[sentiment]
    term_frequency = np.sum(bow, axis=0)
    term_frequencies[sentiment] = term_frequency
    print(f"Sentiment: {sentiment}")
    print(f"Term Frequencies (Before Clamping): {term_frequency}")
    plt.hist(term_frequency, bins=50, edgecolor='black')
    plt.title(f'Term Frequency Distribution (Before Clamping) - {sentiment}')
    plt.xlabel('Frequency')
    plt.ylabel('Count')
    plt.show()
clamped_term_frequencies = {}
for sentiment, term_frequency in term_frequencies.items():
    clamped_term_frequency = np.clip(term_frequency, -LAMBDA, LAMBDA)
```

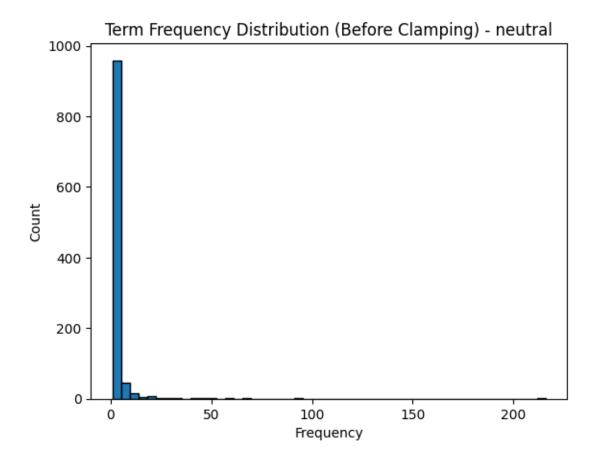
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clamped_term_frequencies[sentiment] = clamped_term_frequency

print(f"Sentiment: {sentiment}")
print(f"Term Frequencies (After Clamping): {clamped_term_frequency}")

#

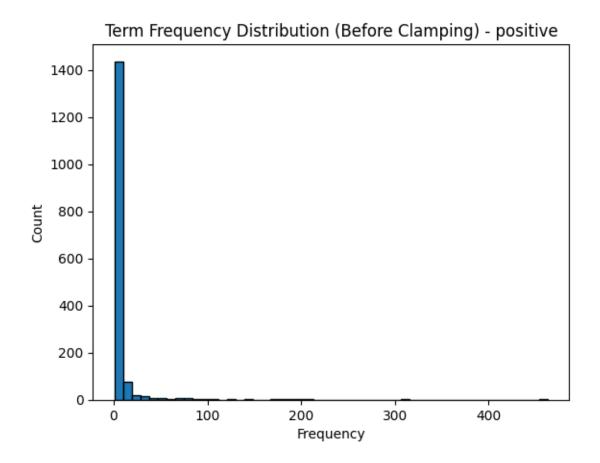
plt.hist(clamped_term_frequency, bins=50, edgecolor='black')
plt.title(f'Term Frequency Distribution (After Clamping) - {sentiment}')
plt.xlabel('Frequency')
plt.ylabel('Count')
plt.show()
```

```
/home/jun/.pyenv/versions/3.11.8/lib/python3.11/site-
packages/torch/cuda/__init__.py:118: UserWarning: CUDA initialization: CUDA
unknown error - this may be due to an incorrectly set up environment, e.g.
changing env variable CUDA_VISIBLE_DEVICES after program start. Setting the
available devices to be zero. (Triggered internally at
../c10/cuda/CUDAFunctions.cpp:108.)
   return torch._C._cuda_getDeviceCount() > 0
/tmp/ipykernel_906294/1462592888.py:47: UserWarning: [W008] Evaluating
Token.similarity based on empty vectors.
   similarities = [(token.similarity(doc), token.lemma_) for token in doc if
token.pos_ in POS]
Sentiment: neutral
Term Frequencies (Before Clamping): [3 1 1 ... 1 1]
```

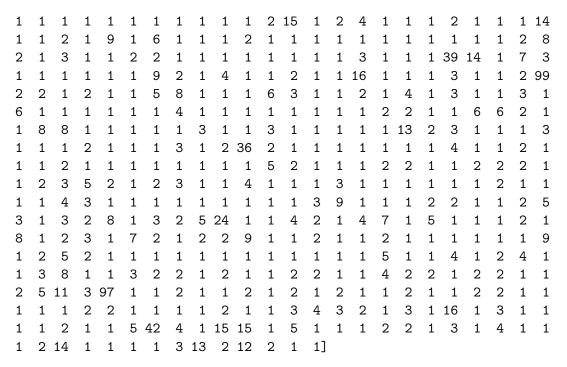


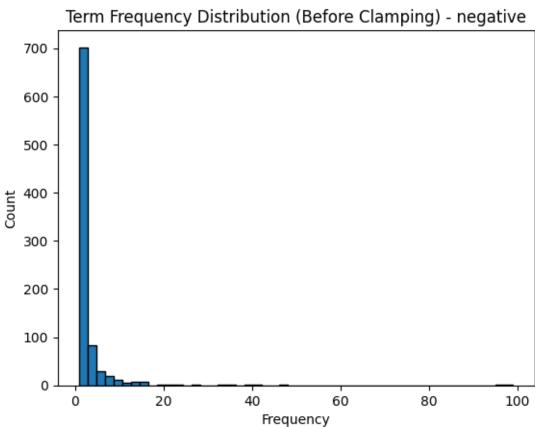
Sentiment: positive

Term Frequencies (Before Clamping): [1 1 1 ... 1 1]



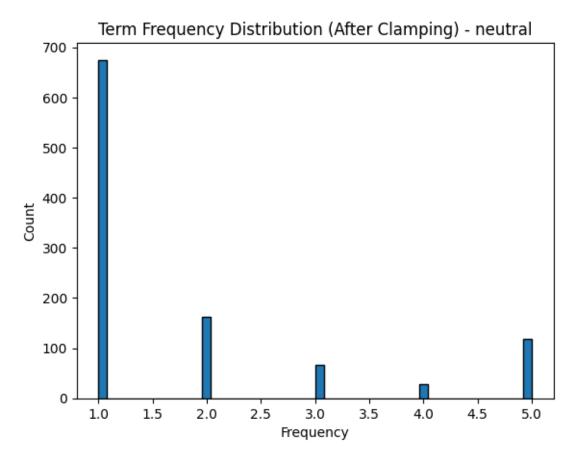
Sentiment: negative																									
Τe	erm	Fre	eque	enc	ies	(Be	efo	re	Cla	amp:	ing):	[1	1	1	1	2	1	1	33	7	1	1	4	1 13
3	3	1	1	1	1	1	1	. 2	! 1	L															
	3	1	1	1	6	1	2	1	1	2	1	1	8	1	16	1	1	1	2	1	1	1	12	1	
	1	2	3	8	5	1	1	27	1	1	10	22	1	2	1	1	1	1	1	2	1	1	1	14	
	1	4	2	1	6	1	1	1	2	2	3	1	1	2	1	10	1	1	6	3	2	2	1	1	
	1	1	2	1	1	1	1	1	1	1	1	2	1	1	1	1	1	1	1	1	6	1	1	1	
	1	3	1	2	1	1	1	1	1	1	1	1	1	1	1	1	2	2	4	1	2	1	1	1	
	6	1	2	1	1	1	1	1	1	4	1	2	3	1	1	1	1	1	1	9	1	1	5	10	
	1	1	2	1	2	1	1	2	1	1	1	1	2	1	3	1	1	2	1	1	1	2	2	1	
	1	4	1	1	1	2	1	2	1	1	1	1	1	2	1	1	1	2	1	1	1	1	7	1	
	1	7	1	1	2	2	1	1	1	1	1	1	1	1	2	1	2	1	1	1	1	6	1	1	
	1	1	6	7	3	3	4	1	1	3	1	1	2	4	1	1	1	4	1	4	4	1	4	6	
	1	1	1	2	1	1	1	1	1	4	1	1	1	1	1	1	1	1	1	1	1	16	1	1	
	1	1	3	1	1	1	1	1	1	5	1	1	1	5	2	6	2	1	1	2	1	1	3	1	
	1	1	1	1	9	3	1	1	1	2	1	4	1	1	3	4	1	3	1	1	1	8	2	8	
	1	1	2	2	1	1	1	1	3	1	1	1	1	1	3	1	1	1	1	1	2	1	1	1	
	1	2	7	1	2	1	8	1	4	1	1	3	48	11	1	1	1	1	1	2	1	4	1	1	
	2	1	1	1	1	1	2	1	1	1	1	2	5	1	2	1	1	1	4	1	1	2	1	1	
	2	1	3	1	2	1	3	1	1	34	1	19	1	1	1	1	1	9	12	1	1	2	1	1	





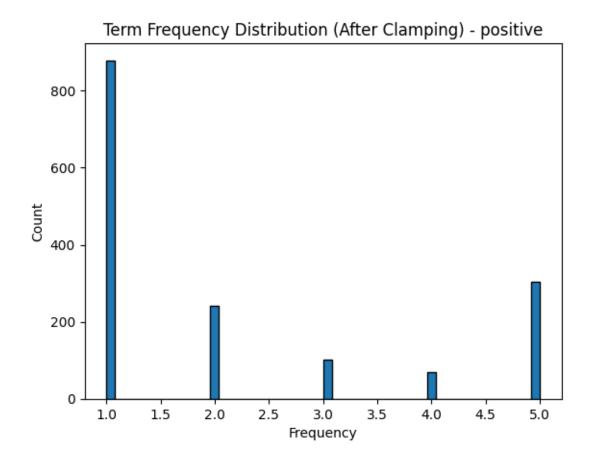
Sentiment: neutral

Term Frequencies (After Clamping): [3 1 1 ... 1 1]



Sentiment: positive

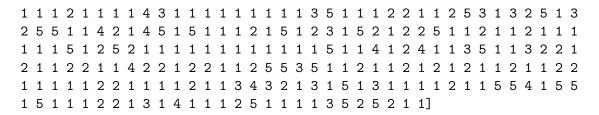
Term Frequencies (After Clamping): [1 1 1 ... 1 1]

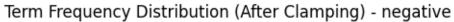


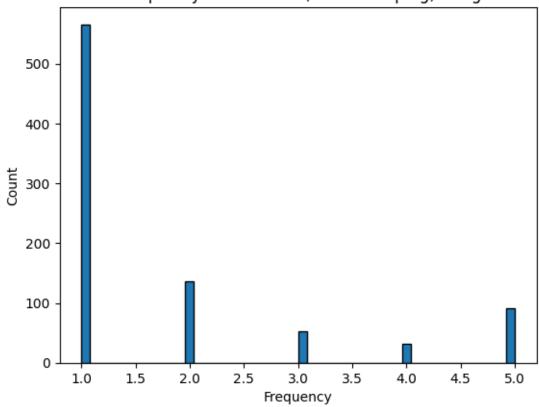
Sentiment: negative

Term Frequencies (After Clamping): [1 1 1 1 2 1 1 5 5 1 1 4 1 5 3 3 1 1 1 1 1 1 2 1 3 1 1 1 5 1 2 1 1 2 1 1 5

 $1\;1\;1\;1\;1\;1\;5\;1\;1\;1\;1\;3\;1\;2\;1\;1\;1\;1\;1\;1\;1\;1\;1\;1\;1\;1\;2\;2\;4\;1\;2\;1\;1\;1\;5\;1\;2\;1$ $1\ 5\ 1\ 1\ 1\ 5\ 2\ 5\ 2\ 1\ 1\ 2\ 1\ 1\ 3\ 1\ 1\ 1\ 1\ 5\ 3\ 1\ 1\ 1\ 2\ 1\ 4\ 1\ 1\ 3\ 4\ 1\ 3\ 1\ 1\ 1$ $1\ 3\ 5\ 5\ 1\ 1\ 1\ 1\ 1\ 2\ 1\ 4\ 1\ 1\ 2\ 1\ 1\ 1\ 1\ 2\ 1\ 1\ 1\ 1\ 2\ 5\ 1\ 2\ 1\ 1\ 1\ 4\ 1\ 1\ 2\ 1$ 151113112522121155111531121 $1\; 5\; 2\; 3\; 1\; 1\; 1\; 3\; 1\; 1\; 1\; 2\; 1\; 1\; 1\; 3\; 1\; 2\; 5\; 2\; 1\; 1\; 1\; 1\; 1\; 1\; 1\; 4\; 1\; 1\; 2\; 1\; 1\; 1\; 2\; 1\; 1$ $1\ 1\ 1\ 1\ 1\ 1\ 5\ 2\ 1\ 1\ 1\ 2\ 2\ 1\ 1\ 2\ 2\ 2\ 1\ 1\ 2\ 3\ 5\ 2\ 1\ 2\ 3\ 1\ 1\ 4\ 1\ 1\ 1\ 3\ 1\ 1\ 1$







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