```
pip install emoji
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Collecting emoji
       Downloading emoji-2.0.0.tar.gz (197 kB)
               | 197 kB 5.1 MB/s
     Building wheels for collected packages: emoji
       Building wheel for emoji (setup.py) ... done
       Created wheel for emoji: filename=emoji-2.0.0-py3-none-any.whl size=193022 sha256=c0230beb5236ec196f6dbd89d226e2241a1
       Stored in directory: /root/.cache/pip/wheels/ec/29/4d/3cfe7452ac7d8d83b1930f8a6205c3c9649b24e80f9029fc38
     Successfully built emoji
     Installing collected packages: emoji
     Successfully installed emoji-2.0.0
pip install contractions
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Collecting contractions
       Downloading contractions-0.1.72-py2.py3-none-any.whl (8.3 kB)
     Collecting textsearch>=0.0.21
       Downloading textsearch-0.0.21-py2.py3-none-any.whl (7.5 kB)
     Collecting pyahocorasick
       Downloading pyahocorasick-1.4.4-cp37-cp37m-manylinux 2 17 x86 64.manylinux2014 x86 64.whl (106 kB)
                                             106 kB 5.2 MB/s
     Collecting anyascii
       Downloading anyascii-0.3.1-py3-none-any.whl (287 kB)
                                           | 287 kB 42.3 MB/s
     Installing collected packages: pyahocorasick, anyascii, textsearch, contractions
     Successfully installed anyascii-0.3.1 contractions-0.1.72 pyahocorasick-1.4.4 textsearch-0.0.21
import pandas as pd
import numpy as np
import emoji
import contractions
import re
from bs4 import BeautifulSoup
import matplotlib.pyplot as plt
```

```
%matplotlib inline
import nltk
import nltk
nltk.download("punkt")
from nltk.tokenize import word tokenize
nltk.download('stopwords')
from nltk.corpus import stopwords
nltk.download('wordnet')
from nltk.stem import LancasterStemmer, WordNetLemmatizer
from nltk import pos tag
from nltk.corpus import wordnet
nltk.download('sentiwordnet')
from sklearn import svm
from sklearn.svm import SVC
from nltk.corpus import sentiwordnet as swn
import pickle
from sklearn.model selection import train test split
from sklearn.feature extraction.text import CountVectorizer
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.model selection import KFold, StratifiedKFold, cross val score
from sklearn.metrics import classification report, confusion matrix, accuracy score
     [nltk data] Downloading package punkt to /root/nltk data...
                   Unzipping tokenizers/punkt.zip.
     [nltk data]
     [nltk data] Downloading package stopwords to /root/nltk data...
                  Unzipping corpora/stopwords.zip.
     [nltk data]
     [nltk_data] Downloading package wordnet to /root/nltk_data...
```

[nltk data] Downloading package sentiwordnet to /root/nltk\_data... Unzipping corpora/sentiwordnet.zip.

# Import Library

[nltk data]

```
df = pd.read_csv('ebay_reviews.csv')
df
```

	category	review title	review content	rating
0	Headsets	Wireless gaming headset	This gaming headset ticks all the boxes # look	5
1	Headsets	Good for those with a big head, low budget	Easy setup, rated for 6 hours battery but mine	3
2	Headsets	MezumiWireless Gaming Headset	I originally bought this wireless headset for	5
3	Headsets	HW- S2 great headset.	This is my 2nd Mezumi headset, It kills the fi	5
4	Headsets	BEST HEADPHONES I'VE PURCHASED IN MY ENTIRE LIFE	This is probably the best headset I've purchas	5
44751	Racks & Holders	Utensil holder	Reasonably priced but a little flimsy	3
44752	Racks & Holders	Recommended	As described	5
44753	Racks &	cheap looking	cheap looking	1

# → Preprocessing

## ▼ 1. Cleaning Data

```
def html_remover(data):
    beauti = BeautifulSoup(data,'html.parser')
    return beauti.get_text()
```

```
def convert_emoji(data):
    return emoji.demojize(data)
def url_remover(data):
    return re.sub('(http|https):\/\\S+', '',data)
def remove round brackets(data):
    return re.sub('\(.*?\)','',data)
def remove punc(data):
    document = re.sub(r'[^\w\s]','', data)
    return document
def white space(data):
    return ' '.join(data.split())
def text_lower(data):
    return data.lower()
def contraction replace(data):
    return contractions.fix(data)
def remove number(data):
    return re.sub(r"\d+", "", data)
def remove singl char(data):
    return re.sub(r"\b[a-zA-Z]\b", "", data)
def web associated(data):
    new data = html remover(data)
    new data = convert emoji(new data)
    new data = url remover(new data)
    new_data = remove_round_brackets(new_data)
    new_data = remove_punc(new_data)
    new_data = white_space(new_data)
    new_data = text_lower(new_data)
    new_data = contraction_replace(new_data)
    new_data = remove_number(new_data)
```

```
new_data = remove_singl_char(new_data)
return new_data

df['cleaning data'] = df['review content'].apply(web_associated)

/usr/local/lib/python3.7/dist-packages/bs4/__init__.py:273: UserWarning: "b'.'" looks like a filename, not markup. You
    ' Beautiful Soup.' % markup)
/usr/local/lib/python3.7/dist-packages/bs4/__init__.py:273: UserWarning: "b'..'" looks like a filename, not markup. You
    ' Beautiful Soup.' % markup)
```

df.head()

cleaning data	rating	review content	review title	category	
this gaming headset ticks all the boxes looks	5	This gaming headset ticks all the boxes # look	Wireless gaming headset	Headsets	0
easy setup rated for hours battery but mine h	3	Easy setup, rated for 6 hours battery but mine	Good for those with a big head, low budget	Headsets	1
originally bought this wireless headset for m	5	I originally bought this wireless headset for	MezumiWireless Gaming Headset	Headsets	2
this is my nd mezumi headset it	-	This is my 2nd Mezumi headset, It	104/00 11 1 1		^

### ▼ 2. Tokenizing

```
def tokenize(data):
    return nltk.word_tokenize(data)
```

```
df['tokenizing'] = df['cleaning data'].apply(tokenize)
```

### ▼ 3. Negation Handling

```
import nltk
nltk.download('omw-1.4')
def Negation(sentence):
    temp = int(0)
    for i in range(len(sentence)):
        if sentence[i-1] in ['not', "n't"]:
            antonyms = []
            for syn in wordnet.synsets(sentence[i]):
                syns = wordnet.synsets(sentence[i])
                w1 = syns[0].name()
                temp = 0
                for 1 in syn.lemmas():
                    if l.antonyms():
                        antonyms.append(1.antonyms()[0].name())
                max dissimilarity = 0
                for ant in antonyms:
                    syns = wordnet.synsets(ant)
                    w2 = syns[0].name()
                    syns = wordnet.synsets(sentence[i])
                    w1 = syns[0].name()
                    word1 = wordnet.synset(w1)
                    word2 = wordnet.synset(w2)
                    if isinstance(word1.wup similarity(word2), float) or isinstance(word1.wup similarity(word2), int):
                        temp = 1 - word1.wup_similarity(word2)
                    if temp>max_dissimilarity:
                        max_dissimilarity = temp
                        antonym_max = ant
                        sentence[i] = antonym_max
                        sentence[i-1] = ''
    while '' in sentence:
```

```
sentence.remove('')
return sentence

[nltk datal Downloading mackage omw-1 4 to /root/nltk data

df['negation'] = df['tokenizing'].apply(Negation)
```

#### Stopword

```
def stopword(data):
    nltk.download('stopwords')
    clean = []
    for i in data:
        if i not in stopwords.words('english'):
            clean.append(i)
    return clean
```

df['stopword'] = df['negation'].apply(stopword)

#### Streaming output truncated to the last 5000 lines.

```
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk data]
            Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
[nltk data]
             Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
```

```
[nltk data]
             Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk_data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
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[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
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[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
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[nltk data] Downloading package stopwords to /root/nltk data...
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             Package stopwords is already up-to-date!
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             Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
[nltk data]
             Package stopwords is already up-to-date!
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[nltk data]
             Package stopwords is already up-to-date!
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             Package stopwords is already up-to-date!
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             Package stopwords is already up-to-date!
[nltk data]
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[nltk data]
             Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
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             Package stopwords is already up-to-date!
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk data...
             Package stopwords is already up-to-date!
[nltk data]
Folto datal Darmlandina madrasa stampanda ta /maat/mltb data
```

#### ▶ 4. Lemmatization

```
[ ] Ļ 3 cells hidden
```

### ▼ 5. Pos Tagging

		category	review title	review content	rating	cleaning data	tokenizing	negation	stopword	lemma	pos_t
	0	Headsets	Wireless gaming headset	This gaming headset ticks all the boxes # look	5	this gaming headset ticks all the boxes looks	[this, gaming, headset, ticks, all, the, boxes	[this, gaming, headset, ticks, all, the, boxes	[gaming, headset, ticks, boxes, looks, grate,	[game, headset, tick, box, look, grate, build,	[(gam Nh (heads NN), (tic NN), (box,
	1	Headsets	Good for those with a big head, low budget	Easy setup, rated for 6 hours battery but mine	3	easy setup rated for hours battery but mine h	[easy, setup, rated, for, hours, battery, but,	[easy, setup, rated, for, hours, battery, but,	[easy, setup, rated, hours, battery, mine, las	[easy, setup, rat, hours, battery, mine, last,	[(easy, J. (setup, NN (rat, NN (hours, N
	2	Headsets	MezumiWireless Gaming Headset	I originally bought this wireless headset for 	5	originally bought this wireless headset for m	[originally, bought, this, wireless, headset,	[originally, bought, this, wireless, headset,	[originally, bought, wireless, headset, xbox,	[originally, buy, wireless, headset, xbox, plu	[(original RB), (bu VE (wireles JJ),
<pre>df.to_csv('preprocessing1.csv', index=False)</pre>											
	વ	Headeete	⊓vv- o∠ great	Мезиті	5	headset it	nu, me∠um,	mezumi	headset	neadset,	NIN
<pre>df = pd.read_csv('preprocessing1.csv')</pre>											

df

		category	review title	review content	rating	cleaning data	tokenizing	negation	stopword	lemma	р
	0	Headsets	Wireless gaming headset	This gaming headset ticks all the boxes # look	5	this gaming headset ticks all the boxes looks	['this', 'gaming', 'headset', 'ticks', 'all',	['this', 'gaming', 'headset', 'ticks', 'all',	['gaming', 'headset', 'ticks', 'boxes', 'looks	['game', 'headset', 'tick', 'box', 'look', 'gr	[( ('he 'NN')
	1	Headsets	Good for those with a big head, low budget	Easy setup, rated for 6 hours battery but mine	3	easy setup rated for hours battery but mine h	['easy', 'setup', 'rated', 'for', 'hours', 'ba	['easy', 'setup', 'rated', 'for', 'hours', 'ba	['easy', 'setup', 'rated', 'hours', 'battery',	['easy', 'setup', 'rat', 'hours', 'battery', '	[('eas <sub>'</sub> ('setup ('rat'
	2	Headsets	MezumiWireless Gaming Headset	I originally bought this wireless headset for	5	originally bought this wireless headset	['originally', 'bought', 'this',	['originally', 'bought', 'this',	['originally', 'bought', 'wireless',	['originally', 'buy', 'wireless',	[('ori( 'RB')
•	→ Pelabelan										
•	TextBlob		HW- S2 great		-		'my', 'nd',	'my', 'nd',		'mezumi',	('m
	[ ] L,5 cel	ls hidden									
•	Vader Se	entiment	IN /F	41 14		41 14	lovahahlul	torokoblu	'best',	'best',	יוסטי/ ייסטי/
	[ ] L,8 cel	ls hidden									

# ▼ Split Data

## ▼ EKSTRAKSI FITUR: Term presence

```
count vect = CountVectorizer(binary=True)
X_train_counts = count_vect.fit_transform(data_label)
print(X_train_counts.shape)
count_vect.vocabulary_
     (44756, 32506)
     {'this': 28335,
      'gaming': 12677,
      'headset': 13937,
      'ticks': 28445,
      'all': 2633,
      'the': 28164,
      'boxes': 4907,
      'looks': 17150,
      'grate': 13269,
      'built': 5221,
      'to': 28587,
      'last': 16517,
      'excellent': 10872,
      'sound': 26303,
      'mic': 18103,
      'comfortable': 6712,
      'wear': 30685,
      'easy': 9927,
      'set': 25191,
      'up': 29734,
      'what': 30830,
      'more': 18586,
      'could': 7552,
      'you': 31594,
```

```
'ask': 3424,
      'for': 12051,
      'setup': 25213,
      'rated': 22793,
      'hours': 14392,
      'battery': 4097,
      'but': 5327,
      'mine': 18233,
      'has': 13798,
      'lasted': 16519,
      'sessions': 25190,
      'over': 20155,
      'good': 13065,
      'loudness': 17213,
      'from': 12338,
      'earcups': 9875,
      'seal': 24921,
      'and': 2896,
      'thick': 28296,
      'padding': 20344,
      'around': 3323,
      'ears': 9903,
      'person': 20864,
      'with': 31098,
      'med': 17872,
      'big': 4457,
      'head': 13903,
      'would': 31276,
      'look': 17139,
      'funny': 12469,
      'on': 19788,
      'small': 25935,
      1 627 21 6404
X_train_TP = count_vect.transform(X_train)
X_test_TP = count_vect.transform(X_test)
print(X_train_TP)
       (0, 2633)
                     1
```

(0, 2896)

1

```
(0, 4097)
              1
(0, 4152)
              1
(0, 4457)
              1
(0, 5031)
              1
(0, 6024)
              1
              1
(0, 7554)
(0, 8221)
              1
(0, 12051)
              1
(0, 12410)
              1
(0, 13065)
              1
              1
(0, 13375)
(0, 13696)
              1
(0, 13737)
              1
(0, 14053)
              1
(0, 15551)
              1
(0, 18333)
              1
(0, 18586)
              1
              1
(0, 18871)
              1
(0, 19216)
(0, 19224)
              1
(0, 21051)
              1
(0, 23893)
              1
(0, 24817)
              1
(35803, 13841)
                       1
(35803, 14872)
                       1
                       1
(35803, 15323)
(35803, 15613)
                       1
(35803, 16876)
                       1
(35803, 16990)
                       1
(35803, 17224)
                       1
(35803, 18954)
                       1
(35803, 19865)
                       1
(35803, 20958)
                       1
(35803, 21929)
                       1
(35803, 24508)
                       1
                       1
(35803, 24683)
(35803, 24974)
                       1
                       1
(35803, 25223)
(35803, 26112)
                       1
(35803, 28164)
                       1
(35803, 28489)
                       1
```

```
    (35803, 28587)
    1

    (35803, 30154)
    1

    (35803, 30850)
    1

    (35803, 31098)
    1

    (35803, 31224)
    1

    (35803, 31259)
    1

    (35803, 31594)
    1
```

## → KLASIFIKASI dengan term presence

```
SVM_Clasifier = svm.SVC(C=1.0, kernel='linear', degree=3, gamma='auto')

# fitting data latih pada classifier
SVM_Clasifier.fit(X_train_TP,y_train)
# memprediksi label pada set data uji
predictions_SVM_TP = SVM_Clasifier.predict(X_test_TP)

# Menggunakan fungsi accuracy_score untuk mendapat nilai akurasi
print('Confusion Matrix: \n',confusion_matrix(y_test, predictions_SVM_TP))
print()
print('Accuracy: ', accuracy_score(y_test, predictions_SVM_TP))
```

#### ▼ EKSTRAKSI FITUR: TF-IDF

```
vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(data_label)
features_train_transformed = vectorizer.transform(X_train)
features_test_transformed = vectorizer.transform(X_test)
print(features_train_transformed)
```

# ▼ KLASIFIKASI dengan TF-IDF

```
SVM Clasifier = svm.SVC(C=1.0, kernel='linear', degree=3, gamma='auto')
mymodel= SVM Clasifier.fit(features train transformed,y train)
#Evaluate the model on the training data set
predictions SVM Tfidf2 = SVM Clasifier.predict(features train transformed)
print(classification report(y train, predictions SVM Tfidf2))
print('Confusion Matrix: \n',confusion matrix(y train, predictions SVM Tfidf2))
print()
print('Accuracy: ', accuracy score(y train, predictions SVM Tfidf2))
#Evaluate the model on the testing data set
predictions SVM Tfidf = SVM Clasifier.predict(features test transformed)
print(classification report(y test, predictions SVM Tfidf))
print('Confusion Matrix: \n',confusion_matrix(y_test, predictions_SVM_Tfidf))
print()
print('Accuracy: ', accuracy_score(y_test, predictions_SVM_Tfidf))
```

### ▼ Evaluation with K-Fold and Classification Report:

```
from sklearn.model_selection import KFold

# vectorizer = TfidfVectorizer()

# X = vectorizer.fit_transform(data_label)
kf = KFold(n_splits=10, shuffle=True, random_state=42)
scores = []
for fold, (train_index, test_index) in enumerate(kf.split(data_label,label), 1):
    X_train, X_test = data_label[train_index], data_label[test_index]
    y_train, y_test = label[train_index], label[test_index]
```

```
vectorizer = TfidfVectorizer()
vectorizer.fit transform(data label)
features train transformed = vectorizer.transform(X train)
features test transformed = vectorizer.transform(X test)
SVM Clasifier = svm.SVC(C=1.0, kernel='linear', degree=3, gamma='auto')
SVM Clasifier.fit(features train transformed,y train)
predictions SVM Tfidf = SVM Clasifier.predict(features test transformed)
print(f'# Fold {fold}, Train set: {len(train index)}, Test set:{len(test index)}')
print(classification report(y test, predictions SVM Tfidf), "\n")
 # Fold 1, Train set: 40280, Test set:4476
               precision
                            recall f1-score
                                               support
     Negative
                    0.89
                              0.36
                                        0.51
                                                    184
      Neutral
                    0.92
                              0.90
                                        0.91
                                                    509
     Positive
                    0.96
                                        0.98
                              0.99
                                                   3783
                                        0.96
                                                   4476
     accuracy
                    0.92
                              0.75
                                        0.80
                                                   4476
    macro avg
                    0.95
                              0.96
                                        0.95
                                                   4476
 weighted avg
 # Fold 2, Train set: 40280, Test set:4476
               precision
                            recall f1-score
                                               support
     Negative
                    0.84
                              0.35
                                        0.50
                                                    217
      Neutral
                    0.92
                              0.88
                                        0.90
                                                    504
     Positive
                    0.96
                              0.99
                                        0.97
                                                   3755
                                        0.95
     accuracy
                                                   4476
                                        0.79
                    0.90
                              0.74
                                                   4476
    macro avg
                    0.95
 weighted avg
                              0.95
                                        0.94
                                                   4476
 # Fold 3, Train set: 40280, Test set:4476
```

support

217

494

recall f1-score

0.49

0.89

0.35

0.89

precision

0.84

0.90

Negative

Neutral

Positive	0.96	0.99	0.97	3765
accuracy			0.95	4476
macro avg	0.90	0.74	0.78	4476
weighted avg	0.94	0.95	0.94	4476
mergineed dig	0.5.	0.33	0.51	1170
# Fold 4, Tra		-		
	precision	recall	f1-score	support
Negative	0.90	0.40	0.56	204
Neutral	0.92	0.90	0.91	494
Positive	0.96	0.99	0.98	3778
accuracy			0.96	4476
macro avg	0.93	0.76	0.81	4476
weighted avg	0.95	0.96	0.95	4476
# Fold 5, Tra	in set: 4028	30, Test s	et:4476	
	precision	recall	f1-score	support
	•			
Negative	0.89	0.45	0.60	214
Neutral	0.93	0.91	0.92	498
Positive	0.96	0.99	0.98	3764
accuracy			0.96	4476
macro avg	0.93	0.78	0.83	4476
weighted avg	0.96	0.96	0.95	4476

Cross-Validation accuracy: nan +/- nan

print('\n\nCross-Validation accuracy: %.3f +/- %.3f' %(np.mean(scores), np.std(scores)))

```
out=out, **kwargs)
/usr/local/lib/python3.7/dist-packages/numpy/core/_methods.py:189: RuntimeWarning: invalid value encountered in double_
  ret = ret.dtype.type(ret / rcount)
/usr/local/lib/python3.7/dist-packages/numpy/core/_methods.py:263: RuntimeWarning: Degrees of freedom <= 0 for slice
  keepdims=keepdims, where=where)
```

/usr/local/lib/python3.7/dist-packages/numpy/core/fromnumeric.py:3441: RuntimeWarning: Mean of empty slice.

/usr/local/lib/python3.7/dist-packages/numpy/core/ methods.py:223: RuntimeWarning: invalid value encountered in true di

```
subok=False)
     /usr/local/lib/python3.7/dist-packages/numpy/core/ methods.py:254: RuntimeWarning: invalid value encountered in double
       ret = ret.dtype.type(ret / rcount)
review input transformed = vectorizer.transform(data label)
SVM prediction = SVM Clasifier.predict(review input transformed)
import collections, numpy
print("Jumlah review: {}".format(len(data label)))
SVM = collections.Counter(predictions SVM Tfidf)
print("Hasil Klasifikasi SVM : ", SVM)
results = pd.DataFrame({
    "Labeled_Data" : data_label,
    "Label" : SVM prediction
 })
results.to_csv("Hasil_SVM2.csv", index = False)
     Jumlah review: 44756
     Hasil Klasifikasi SVM : Counter({'Positive': 3899, 'Neutral': 492, 'Negative': 84})
X_test.shape
```

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