```
from google.colab import drive
drive.mount('/content/drive')
     Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.m
%cd '/content/drive/MyDrive/Studi Independent/NLP/'
     /content/drive/MyDrive/Studi Independent/NLP
import numpy as np
import pandas as pd
df=pd.read_csv('IMDB Dataset.csv')
df.head()
df.describe()
df.isna().sum()
     review
                  0
     sentiment
                  0
     dtype: int64
```

Data Cleaning

import nltk

```
from sklearn.feature extraction.text import CountVectorizer
from nltk.corpus import stopwords
from wordcloud import WordCloud, STOPWORDS
from nltk.stem import WordNetLemmatizer, LancasterStemmer
from nltk.tokenize import word_tokenize
from nltk.tokenize.toktok import ToktokTokenizer
import re, string, unicodedata
from string import punctuation
df['review']
     0
              One of the other reviewers has mentioned that ...
     1
              A wonderful little production. <br /><br />The...
              I thought this was a wonderful way to spend ti...
     3
              Basically there's a family where a little boy ...
              Petter Mattei's "Love in the Time of Money" is...
              I thought this movie did a down right good job...
     49995
     49996
              Bad plot, bad dialogue, bad acting, idiotic di...
     49997
              I am a Catholic taught in parochial elementary...
              I'm going to have to disagree with the previou...
     49998
     49999
              No one expects the Star Trek movies to be high...
     Name: review, Length: 50000, dtype: object
def hapus_kurung(text):
  return re.sub('\[[^]]*\]', '', text)
def hapus_url(text):
  return re.sub(r'https\S+', '', text)
def hapus spesial character(text, remove digits=True):
  pattern = r'[^a-zA-z0-9\s]'
  text = re.sub(pattern, '', text)
  return text
nltk.download('stopwords')
stop = set(stopwords.words('english'))
punctuation = list(string.punctuation)
stop.update(punctuation)
def remove stopwords(text):
  final text = []
  for i in text.split():
    if i.strip().lower() not in stop and i.strip().lower().isalpha():
      final text.append(i.strip().lower())
  return " ".join(final text)
     [nltk data] Downloading package stopwords to /root/nltk data...
                 Package stopwords is already up-to-date!
     [nltk data]
def clean_text(text):
  text = hapus kurung(text)
  text = hapus url(text)
  text = hapus_spesial_character(text, True)
  text = remove stopwords(text)
  return text
```

 \Box

```
df['review'] = df['review'].apply(clean_text)
df.head()
```

sentiment	review	
positive	one reviewers mentioned watching oz episode yo	0
positive	wonderful little production br br filming tech	1
positive	thought wonderful way spend time hot summer we	2
negative	basically theres family little boy jake thinks	3
positive	petter matteis love time money visually stunni	4

```
df.sentiment.replace("positive", 1, inplace=True)
df.sentiment.replace("negative", 0, inplace=True)
df.head()
```

→ Membuat Wordcloud

```
import matplotlib.pyplot as plt

plt.figure(figsize = (20,20))
wc = WordCloud(max_words = 2000, width = 1600, height=800).generate(" ".join(df[df.sentime plt.imshow(wc, interpolation = 'bilinear')
```

```
text = df['review'].values.tolist()
label = df['sentiment'].values
```

Vektorisasi

accuracy

```
from \ sklearn. feature\_extraction. text \ import \ Count Vectorizer, \ Tfidf Vectorizer
tfidf_vectorizer = TfidfVectorizer().fit(text)
tfidf_text=tfidf_vectorizer.transform(text)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(tfidf_text, label, test_size = 0.3, ra
from sklearn.svm import SVC
from sklearn.metrics import classification_report
model = SVC().fit(X_train, y_train)
pred_svm = model.predict(X_test)
print(classification_report(pred_svm, y_test))
                   precision
                              recall f1-score
                                                    support
                                   0.91
                                             0.90
                        0.88
                                                       7248
                        0.91
                                   0.89
                                             0.90
                                                       7752
```

0.90

15000

macro avg 0.90 0.90 0.90 15000 weighted avg 0.90 0.90 0.90 15000

×