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## LAMPIRAN 1

### HASIL PENGUJIAN

No	Tweet	Hasil Uji
1	habis bapaknya bahas tunjangan nakes bidan rumah sakit pemerintah daerah gajinya dikurangi akibat covid anggaran dialihkan apd apd dibatasi rumah sakit	Sentiment Positif
2	bergantian sentimen positif negatif menyambang bursa negeri global tekanan covid fluktuasinya harga komoditas stimulus pemerintah katalis jaga indeks rentang 4 minggu sideways	Sentiment Positif
3	jujur beliau perihal covid 19 nih sekeluarga teman teman kesal pasalnya kebijakan menghambat penularan diprotes presiden gubernur daerah khusus kota dilanggar giliran kena pemerintah disalahkan	Sentiment Positif
4	berita sonora presiden joko widodo segan segan mengeluarkan peraturan pemerintah pengganti undang undang peraturan pemerintah pengganti undang undang mempermudah kinerja menteri merealisasikan program krisis akibat pandemi covid 19 via tribunnews	Sentiment Positif
5	pakai duit duit pemerintah mbak biayain pemerintah kema covid	Sentiment Positif
6	presiden joko widodo jokowi segan segan mengeluarkan peraturan pemerintah pengganti undang undang peraturan pemerintah pengganti undang undang mempermudah kinerja menteri merealisasikan program krisis akibat pandemi covid 19	Sentiment Positif
7	kompetisi liga 1 liga 2 liga 3 oktober 2020 memperhatikan ketentuan protokol kesehatan covid 19 ditetapkan pemerintah	Sentiment Positif

8	untungnya bilang kadrun ajak demo pakai baju putih duduk pojokan bawa nasi bungkus drun skill kadrun musim covid pemutusan hubungan kerja eksis dipecat kadrun kayak km iya job sepi demo dilarang pemerintah	Sentiment Negatif
9	lambene sok sarjana internet bebas bro orang berbicara apapun forum diskusi ilmiah jurusan teknik mesin iya karep2ku bro penanganan covid 19 pemerintah buruk logika mas	Sentiment Negatif
10	logika nya 1 kejadian wuhan pemerintah bergerak lambat menjemput warga negara indonesia wuhan 2 dunia menutup bandara menghambat penyebaran covid indonesia membayar buzzer pariwisata	Sentiment Negatif
11	presiden jokowi bertambahnya pasien covid 19 indikasi daerah keuntungan finansial ditetapkan pasien penyakit pasien covid 19 dana pemerintah mengucur deras kspgoid	Sentiment Positif
12	kolega hcm hanoi point utamanya pemerintah menganggap covid 19 serius diawal2 wuhan keseriusan ditularkan masyarakat kedai kopi buka dikirim fotonya	Sentiment Positif

## LAMPIRAN 2

### LISTING PROGRAM

```
# Downloading some useful nltk libraries

import nltk
nltk.download('punkt')
nltk.download('stopwords') # to tokenize words and remove words <3

# Importing Libraries

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import style
style.use('ggplot')
import seaborn as sns
import re
from textblob import TextBlob
from wordcloud import WordCloud
from nltk.tokenize import word_tokenize
from nltk.stem import PorterStemmer
from nltk.corpus import stopwords
stop_words = set(stopwords.words('english'))
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix, ConfusionMatrixDisplay
import warnings
warnings.filterwarnings('ignore')

dataset = pd.read_excel('datacovid.xlsx')
dataset.head()

dataset.info() # getting info about the dataset
```

```
newdataset = dataset.drop(['date', 'time', 'username'], axis=1)
newdataset.head()
```

```
def textprocessing(text):
    text = text.lower()

    text = re.sub(r"https\S+|www\S+https\S+", " ", text, flags=re.MULTILINE)
    text = re.sub(r"@w+|\#", " ", text)

    text = re.sub(r"[^\w\s]", " ", text)

    text_tokens = word_tokenize(text)
    filtered_text = [w for w in text_tokens if not w in stop_words]
    return " ".join(filtered_text)
```

```
newdataset.text = dataset['tweet'].apply(textprocessing)
```

```
newdataset = newdataset.drop_duplicates('tweet')
```

```
stemmer = PorterStemmer()
def stemming(data):
    text = [stemmer.stem(word) for word in data]
    return data
```

```
newdataset['tweet'] = newdataset['tweet'].apply(lambda x : stemming(x))
```

```
# getting polarity of the text
```

```
def polarity(text):
    return TextBlob(text).sentiment.polarity
```

```
# getting subjectivity of the text
```

```

def subjectivity(text):
    return TextBlob(text).sentiment.subjectivity

# creating a polarity column for the text
# creating a subjectivity column for the text

newdataset['polarity'] = newdataset['tweet'].apply(polarity)
newdataset['subjectivity'] = newdataset['tweet'].apply(subjectivity)

# viewing updated dataset

newdataset.head()

# calculating sentiment of the data using polarity values

def sentiment(label):
    if label < 0 :
        return "Negative"
    elif label == 0 :
        return "Neutral"
    else :
        return "Positive"

newdataset['Sentiment'] = newdataset['polarity'].apply(
sentiment) # applying Sentiment column to newdataset
which analyzes polarity values of the text

# viewing the dataset

newdataset.head()

# plotting the Sentiments in a bargraph

pic = plt.figure(figsize=(5,5))
sns.countplot(x='Sentiment',data=newdataset)

# plotting polarity and subjectivity

```

```

'''plt.figure(figsize=(5,5))

for i in range(0,newdataset.shape[0]):
    plt.scatter(newdataset['polarity'][i],newdataset['subjectivity'][i],color = 'blue')

plt.title("Sentiment Analysis")
plt.xlabel("Polarity")
plt.ylabel("Subjectivity")
plt.show()'''

# viewing positive tweets

positive_tweets = newdataset[newdataset.Sentiment == "Positive"]
positive_tweets = positive_tweets.sort_values(['polarity'],ascending=False)
positive_tweets.head()

# viewing positive tweets in a wordcloud

Content = " ".join([word for word in positive_tweets['tweet']])
plt.figure(figsize=(20,20),facecolor="None")
wordcloud = WordCloud(max_words=500,width=1500,height=1000).generate(Content)
plt.imshow(wordcloud,interpolation="bilinear")
plt.axis("off")
plt.title("Frequent words in Positive Tweets")
plt.show()

# viewing negative tweets

negative_tweets = newdataset[newdataset.Sentiment == "Negative"]
negative_tweets = negative_tweets.sort_values(['polarity'],ascending=False)
negative_tweets.head()

# viewing negative tweets in a wordcloud

```



```
Content = " ".join([word for word in negative_tweets['tweet']])
plt.figure(figsize=(20,20),facecolor="None")
wordcloud = WordCloud(max_words=500,width=1500,height=1000).generate(Content)
plt.imshow(wordcloud,interpolation="bilinear")
plt.axis("off")
plt.title("Frequent words in Negative Tweets")
plt.show()
```

```
# viewing neutral tweets
```

```
neutral_tweets = newdataset[newdataset.Sentiment == "Neutral"]
neutral_tweets = neutral_tweets.sort_values(['polarity'],ascending=False)
neutral_tweets.head()
```

```
# viewing neutral tweets in a wordcloud
```

```
Content = " ".join([word for word in neutral_tweets['tweet']])
plt.figure(figsize=(20,20),facecolor="None")
wordcloud = WordCloud(max_words=500,width=1500,height=1000).generate(Content)
plt.imshow(wordcloud,interpolation="bilinear")
plt.axis("off")
plt.title("Frequent words in Neutral Tweets")
plt.show()
```

```
# creating a SVM Model
```

```
# for features in Support Vector Classification
```

```
vect = CountVectorizer(ngram_range=(1,2)).fit(newdataset['tweet'])
```

```
# creating feature names
```

```
feature_names = vect.get_feature_names()
print("Number of features are :",len(feature_names))
```

```

print("First 20 Features are :",feature_names[:20])

X = newdataset['tweet']
Y = newdataset['Sentiment']
X = vect.transform(X)

x_train , x_test , y_train ,y_test = train_test_split(X
,Y,test_size=0.1,random_state=20)

print("Size of x_train :",x_train.shape)
print("Size of y_train :",y_train.shape)
print("Size of x_test :",x_test.shape)
print("Size of y_test :",y_test.shape)

# implementing SupportVectorMachine using define Support
tVectorClassification function

from sklearn.svm import SVC
reg = SVC(kernel = 'linear', random_state = 0)
reg.fit(x_train,y_train)
y_pred = reg.predict(x_test)
y_acc = accuracy_score(y_pred,y_test)
print("Test Accuracy : ",y_acc*100)

# Printing Confusion Matrix

print(confusion_matrix(y_test,y_pred))
print()
print(classification_report(y_test,y_pred))

# Another form of Confusion Matrix

style.use('classic')
cm = confusion_matrix(y_test,y_pred,labels=reg.classes_
)
display = ConfusionMatrixDisplay(confusion_matrix = cm,
display_labels=reg.classes_)
display.plot()

```

## LAMPIRAN 3

### DAFTAR RIWAYAT HIDUP

#### (*CURRICULUM VITAE*)

#### 1. DATA PRIBADI

Nama : Rafizah Aini  
Nim : 0701162014  
Tempat, Tanggal Lahir : Medan, 23 Juni 1998  
Alamat : Jl. Cirebon No.1 LK XXIV  
Kel/Desa : Belawan II  
Kecamatan : Medan Belawan  
Kabupaten : Kota Medan  
Agama : Islam  
Status : Belum Menikah  
Nama Orang Tua  
Ayah : Rahmadsyah Putra  
Ibu : Chadijah



#### 2. DATA PENDIDIKAN

SD : MIN Belawan  
SMP : SMP Hang Tuah 1 Belawan  
SMA : SMA Hang Tuah Belawan  
Perguruan Tinggi : Universitas Islam Negeri Sumatera Utara

LAMPIRAN 4









KARTU BIMBINGAN SKRIPSI

KARTU BIMBINGAN SKRIPSI

Semester Gasal/Genap Tahun Akademik ...../.....

Nama : Rafizah Aini	Pembimbing I : Dr. Mhd. Furgan, S.Si, M. Comp. Sc
NIM : 0901162014	Pembimbing II : Muhammad Ikhsan, ST, M. Kom.
Prog. Studi : Ilmu Komputer	SK Pembimbing :
Judul Skripsi : Analisis Sentimen Opini Masyarakat Indonesia Terhadap Pandemi Virus Corona (Covid-19) Di Media Sosial Twitter Menggunakan Metode Support Vector Machine	

P E R T	PEMBIMBING I			PEMBIMBING II		
	Tgl.	Materi Bimbingan	Tanda Tangan	Tgl.	Materi Bimbingan	Tanda Tangan
I	23 Nov 2020	Bimbingan Bab I, II dan III		23 Nov 2020	Bimbingan Bab I, II, III	
II	7 Juni 2022	Revisi Bab II		2 Juni 2022	Revisi Penulisan	
III	13 Juni 2022	Revisi Bab III		14 Juni 2022	Revisi Bab II dan III	
IV	23 Juni 2022	Acc Seminar Proposal		23 Juni 2022	Acc Seminar Proposal	
V	12 Des 2022	Revisi Bab IV dan V		19 Des 2022	Revisi Bab IV dan V	

VI	16 Jan 2023	Revisi Bab IV		23 Jan 2023	Revisi Bab IV	
VII	30 Jan 2023	Revisi Program		25 Jan 2023	Revisi Program	
VIII	7 Feb 2023	Acc Sidang Munafasyah		8 Feb 2023	Acc Sidang Munafasyah	
IX						
X						

Medan, ..... 2023.

An. Dekan

Ketua Jurusan/Program Studi

Ilmu Komputer

  
ILK. ZUFRIA, M. Kom

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Catatan: Pada saat bimbingan, kartu ini harus diisi dan ditandatangani oleh pembimbing