

## LAMPIRAN

## **Lampiran 1 – Daftar Riwayat Hidup**

Data Pribadi

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## Riwayat Pendidikan

Periode (Tahun)	Sekolah/Institusi	Jurusan	Jenjang Pendidikan
2007-2013	SDN 11 Pagi Joglo	-	SD
2013-2016	SMPN 206 Jakarta Barat	-	SMP
2016-2019	SMA 101 Jakarta Barat	MIPA	SMA

## Lampiran 2 – Kode Untuk *Preprocessing*

```
In [1]: !pip install -U scikit-learn  
!pip3 install sklearn  
!pip install nltk  
!pip install openpyxl  
!pip install pandas  
!pip install Sastrawi  
import string  
from sklearn.pipeline import Pipeline  
import pandas as pd  
import numpy as np  
import re  
import nltk
```

```
In [2]: data = pd.read_excel ('Data_Tweet.xlsx')
data.head(2000)
```

	username	Text
0	@aetheriusjk	lagi di fase frustasi, paling bentar lagi depresi
1	@youngbutcher94	RT @karminnas_ _Ini orang yang dulu bilang kusu...
2	@xicty	RT @TraseerNew: @fromureyes (@tanyantes @tanyantes ...
3	@makanniyaindomie	depresi setelah belajar mtk minato? Yippeee! https://...
4	@kkobopb	Akhinya, terimakasih sudah membantu aku, kaak...
...	...	...
1497	@ayupurwatie	Hebat yah! Diluar bagaimana keadaan baby, dan ...
1498	@bbmyao	@megworeng dulu kakak aku langsung jadi sidi, f...
1499	@Legister	@leadsHalmark Mau tukar poin jadi kesehatan me...
1500	@llisllis	Mencari keterungan batin dan kesehatan mental...
1501	@_riverheaven	RT @_riverheaven: Demi menjaga kesehatan menta...

```
In [3]: df = data

In [4]: #Tahap Casefolding

def casefolding(kalimat):
    # Mengubah huruf kapital menjadi huruf kecil
    kalimat = str(kalimat).lower()
    # Menghapus username Twitter (yang dimulai dengan @)
    kalimat = re.sub("@\w+", " ", kalimat)

    #menghapus hastag
    kalimat = re.sub(r"^\#\w+", "", kalimat)

    #Menghapus Url
    kalimat = re.sub(r"http\S+|www\S+|https\S+", "", kalimat)

    # Menghapus karakter non-alphanumeric, kecuali @ dan #
    kalimat = re.sub(r"^[^a-zA-Z0-9@#]", " ", kalimat)

    #menghapus spasi
    kalimat = kalimat.strip()

    # Menghapus angka menggunakan regular expression
    kalimat = re.sub(r'\d+', ' ', kalimat)

    return kalimat

In [5]: df['Casefolding'] = df['Text'].apply(casefolding)
df.head(1500)

In [6]: from nltk.tokenize import word_tokenize

In [16]: #Tahap Tokenizing

def tokenize_text(text):
    tokens = nltk.tokenize.word_tokenize(text)
    return tokens

In [8]: df['Token'] = df['Casefolding'].apply(tokenize_text)
df.head(200)

In [9]: #Stopwords
from Sastrawi.StopWordRemover.StopWordRemoverFactory import StopWordRemoverFactory
Factory = StopwordRemoverFactory()
stopwords = Factory.get_stop_words()
print (stopwords)

['yang', 'untuk', 'pada', 'ke', 'para', 'namun', 'menurut', 'antara', 'dia', 'dua', 'ia', 'seperti', 'jika', 'jika', 'sehingga', 'kembali', 'dan', 'tidak', 'ini', 'karena', 'kepada', 'oleh', 'saat', 'harus', 'sementara', 'setelah', 'belum', 'kami', 'sekitar', 'bagi', 'serta', 'dengan', 'akan', 'juga', 'ada', 'mereka', 'sudah', 'saya', 'terhadap', 'se cara', 'agar', 'lain', 'andalah', 'itu', 'dalam', 'biasa', 'bahwa', 'atau', 'hanya', 'kita', 'dengan', 'akan', 'juga', 'ada', 'mereka', 'sudah', 'saya', 'terhadap', 'se cara', 'agar', 'lain', 'and', 'begitu', 'mengapa', 'kenapa', 'yaitu', 'yakin', 'dari pada', 'titulah', 'lagi', 'maka', 'tentang', 'demi', 'dimana', 'keman', 'pula', 'sambil', 'sebelum', 'sesudah', 'supaya', 'guna', 'kah', 'pun', 'sampai', 'sedangkan', 'selagi', 'sementara', 'teta', 'pi', 'apakah', 'kecuali', 'sebab', 'selain', 'seolah', 'seraya', 'seterusnya', 'tanpa', 'agak', 'boleh', 'dapat', 'dsb', 'dst', 'dll', 'dahulu', 'dulunya', 'anu', 'demikian', 'tapi', 'ingin', 'juga', 'nggak', 'mari', 'nanti', 'melainkan', 'oh', 'ok', 'seh', 'arunya', 'sebutulnya', 'setiap', 'setidaknya', 'sesuatu', 'pasti', 'saja', 'toh', 'ya', 'walau', 'tolong', 'tentu', 'amat', 'a', 'palagi', 'bagaimanapun']

In [10]: def stopword_text(tokens):
    cleaned_tokens = []
    for token in tokens:
        if token not in stopwords:
            cleaned_tokens.append(token)
    return cleaned_tokens

In [11]: df['Stopwords'] = df['Token'].apply(stopword_text)
df.head(1000)

In [12]: #Proses Stopword Extend
from nltk.corpus import stopwords
nltk.download("stopwords")
list_stopwords = stopwords.words('indonesian', 'english')

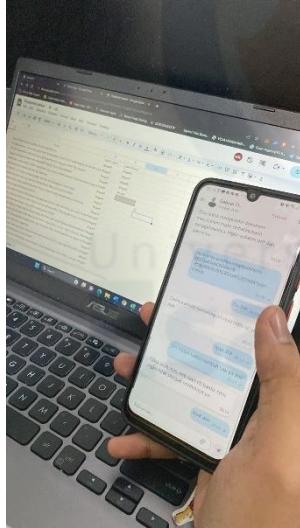
list_stopwords.extend(["stlh","xf0","x9f","xa4","xa4","part","sih","deact","rt","pls","cut","off","xf0","x9f","x99","x8f",
"bentar","sd","kahdhevjbswkwhishhw","i","pasc","#","ahhhh","t","ya","walach","abiez","time","true",
"nya","wukwu","happy","ending","you","should","know","how","much","i","love","you","my",
"baby","my","proud","boy","itadori","yuuji","what","s","wrong","with","me","9","10","y","2",
"2ny","triggered","co","hayang","bunuh","alesan","ung","kamamana","mawa","peso",
"karna","neangan","psnskd","wkzn","w","amp","fc","royal","gk","holaaa","wuku",
"anjg","goblokzin","sifl","harm","nk","gi","ic","exo","jombo","sa","tai","oink",
"anj","pasca","konser","hrgainki","kontol","bajingan","this","is","trying","cloe",
"ainkkkkkkk","produce","bitch","encore","tds","clash","royale","anjggg","panal",
"kop","xe","x80","xa6","n","nhaw","play","dil","doi","bmr","pens","ipuel","etmin",
"waiji","bola","kimak","fyp","angst","tolol","jkt48","buddies","halo","kemenkes","1500",
"567","marshana","selesaikan","charger","ipad","ipadnya","gem","batre","x98","x8c","xa5",
"xb2","cos","jakarta","bell","maskot","lucu","kim","junyeon","misstreatment","im","so","of",
"him","fucked","up","solo","netflixnya","soc","an","anjingg","anying","cok","ajg","anjg",
"anjing","ace","fr","sing","kontol","ong","ygv","qrtsnsya","asfdgsh","haechan",
"anjrit","akut","awikukok","lol","setan","asu","dil","co","ajgga","wuk","tai","wukuk",
"sayanakk","anhing","bacot","hts","hragimaki","cegil","cogil","e","amu","jet","piye",
"maneh","nyet","bgct","pens","ipuel","kimak","fyp","karehsiusiguru","rollercoaster","ucl",
"xd","rengginang","dragon","ball","bola","ajggg88","anyinggg","idol","yysf","qnt","anjenggg",
"goblok","apple","taskimalaya","lo","yg","main","art","oliver","xf","x97","xb","uu","lt",
"bod","humppannya","xc","it","not","dr","jina","ardi","sp","kj","tu","ket","anjing","pjms","jm",
"m","ah","mo","alywafa","dot","con","b","a","bomandir","ken","adhd","haha","hiii","th","si",
"aaaaaa","tu","hadeu","ama","ni","kekukuk","tibab","taraaaaa","ekukukuk","wlukukuk","gin","hehehe",
"wow","pl","ed","bangsal","mmc","gin","plijjj","snsjuj","am","meuehkkk","gin","ken","iku","cukkk",
"gwaa","yessssss","kek","ti","heneheneheneh","p","ouch","aka","bon","hahihii","wakakaka","bjin",
"anjing","nanaha","em","koar","jog","v","yoo","ren","lgi","c","wkw","hehe","hnnn","ber","awokwok","ciu",
"anjir","yaah","si","anirrrr","awokwok","bosskuu","rojeh","blekping","imo","to","yo","yongi","pd",
"riks","lim","ff","dlu","tpi","jdi","huuh","cc","tello","ame","dog","ansharparam","ret","msi","okaaay",
"wukukuk","ber","uras","ki","brantem","selamaa","tp","wehhhh","sia","apasihhhhhhhhh","k","youtube",
"watch","qldtlvfy","ab","bgst","nb","dar","wkwkw","h","gt","namjoon","un","x","f","nnarahubung","yeuhht",
"yh"])

list_stopwords = set(list_stopwords)

def stopwords_removal(text):
    return [word for word in text if word not in list_stopwords]

df['Stopwords_new'] = df['Stopwords'].apply(stopwords_removal)
df.head(1000)
```

## Lampiran 3 – Pelabelan Yang Dilakukan Oleh Pakar



Merupakan proses pelabelan validasi oleh pakar yang didapat melalui halodoc.

## Lampiran 4 – Algoritma *Naïve Bayes*

```
In [38]: import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score, f1_score
import seaborn as sns
import matplotlib.pyplot as plt

# Step 1: Import Library and read the dataset
data_NB = pd.read_csv("Label_Cleaned.csv")

# Step 2: Preprocessing (if needed) - Not shown in this example as it depends on the data

# Step 3: Split the data into features (tweets) and Labels (Labels)
tweets = data_NB['Tweet']
labels = data_NB['Label']

# Step 4: Feature extraction using TF-IDF
tfidf_vectorizer = TfidfVectorizer(max_features=1000) # You can adjust the number of features as needed
X_tfidf = tfidf_vectorizer.fit_transform(tweets)

# Step 5: Split the data into training and testing sets (80% - 20%)
X_train, X_test, y_train, y_test = train_test_split(X_tfidf, labels, test_size=0.2, random_state=42)

# Membuat objek Naive Bayes
nb_classifier = MultinomialNB()

# Melatih model dengan data pelatihan
nb_classifier.fit(X_train, y_train)

# Melakukan prediksi pada data pengujian
y_pred = nb_classifier.predict(X_test)

# Step 6: Calculate metrics - Confusion matrix, accuracy, precision, recall, and f1 score
confusion_mat = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='weighted')
recall = recall_score(y_test, y_pred, average='weighted')
f1 = f1_score(y_test, y_pred, average='weighted')

# Convert metrics to percentage and round to 2 decimal places
accuracy_percent = round(accuracy * 100, 2)
precision_percent = round(precision * 100, 2)
recall_percent = round(recall * 100, 2)
f1_percent = round(f1 * 100, 2)

print("Confusion Matrix:")
print(confusion_mat)

# Step 7: Plot the confusion matrix as a heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(confusion_mat, annot=True, fmt="d", cmap="Blues", xticklabels=["Negatif", "Positif"], yticklabels=["Negatif", "Positif"])
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix")
plt.show()

print("")
print("== Hasil Evaluasi Naive Bayes Classifier ==")
print("Accuracy: {}%".format(accuracy_percent))
print("Precision: {}%".format(precision_percent))
print("Recall: {}%".format(recall_percent))
print("F1 Score: {}%".format(f1_percent))
```

## Lampiran 5 – Algoritma Support Vector Machine

```
In [39]: import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score, f1_score
import seaborn as sns
import matplotlib.pyplot as plt

# Step 1: Import Library and read the dataset
data_SV = pd.read_csv("Label_Cleaned.csv")

# Step 2: Preprocessing (if needed) - Not shown in this example as it depends on the data

# Step 3: Split the data into features (tweets) and Labels (Labels)
tweets = data_SV['Tweet']
labels = data_SV['Label']

# Step 4: Feature extraction using TF-IDF
tfidf_vectorizer = TfidfVectorizer(max_features=1000) # You can adjust the number of features as needed
X_tfidf = tfidf_vectorizer.fit_transform(tweets)

# Step 5: Split the data into training and testing sets (80% - 20%)
X_train, X_test, y_train, y_test = train_test_split(X_tfidf, labels, test_size=0.2, random_state=42)

# Membuat objek SVM
svm_classifier = SVC(kernel='linear', random_state=42)

# Melatih model dengan data pelatihan
svm_classifier.fit(X_train, y_train)

# Melakukan prediksi pada data pengujian
y_pred = svm_classifier.predict(X_test)

# Step 6: Calculate metrics - Confusion matrix, accuracy, precision, recall, and f1 score
confusion_mat = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='weighted')
recall = recall_score(y_test, y_pred, average='weighted')
f1 = f1_score(y_test, y_pred, average='weighted')

# Calculate metrics in percentage
accuracy_percent = accuracy * 100
precision_percent = precision * 100
recall_percent = recall * 100
f1_percent = f1 * 100

print("Confusion Matrix:")
print(confusion_mat)

# Step 7: Plot the confusion matrix as a heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(confusion_mat, annot=True, fmt="d", cmap="Greens", xticklabels=["Negatif", "Positif"], yticklabels=["Negatif", "Positif"])
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix")
plt.show()

print("")
print("== Hasil Evaluasi Support Vector Machine ==")
print("Accuracy:", accuracy_percent, "%")
print("Precision:", precision_percent, "%")
print("Recall:", recall_percent, "%")
print("F1 Score:", f1_percent, "%")
```

## Lampiran 6 – Algoritma Random Forest

```
In [48]: import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score, f1_score
import seaborn as sns
import matplotlib.pyplot as plt

# Step 1: Import Library and read the dataset
data_RF = pd.read_csv("Label_Cleaned.csv")

# Step 2: Preprocessing (if needed) - Not shown in this example as it depends on the data

# Step 3: Split the data into features (tweets) and labels (Labels)
tweets = data_RF['Tweet']
labels = data_RF['Label']

# Step 4: Feature extraction using TF-IDF
tfidf_vectorizer = TfidfVectorizer(max_features=1000) # You can adjust the number of features as needed
X_tfidf = tfidf_vectorizer.fit_transform(tweets)

# Step 5: Split the data into training and testing sets (80% - 20%)
X_train, X_test, y_train, y_test = train_test_split(X_tfidf, labels, test_size=0.2, random_state=42)

# Membuat objek Random Forest Classifier
rf_classifier = RandomForestClassifier(random_state=42)

# Melatih model dengan data pelatihan
rf_classifier.fit(X_train, y_train)

# Melakukan prediksi pada data pengujian
y_pred = rf_classifier.predict(X_test)

# Step 6: Calculate metrics - Confusion matrix, accuracy, precision, recall, and f1 score
confusion_mat = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='weighted')
recall = recall_score(y_test, y_pred, average='weighted')
f1 = f1_score(y_test, y_pred, average='weighted')

# Convert metrics to percentage and round to 2 decimal places
accuracy_percent = round(accuracy * 100, 2)
precision_percent = round(precision * 100, 2)
recall_percent = round(recall * 100, 2)
f1_percent = round(f1 * 100, 2)

print("Confusion Matrix:")
print(confusion_mat)

# Step 7: Plot the confusion matrix as a heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(confusion_mat, annot=True, fmt="d", cmap="Oranges", xticklabels=["Negatif", "Positif"], yticklabels=["Negatif", "Positif"])
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix")
plt.show()

print("== Hasil Evaluasi Random Forest ==")
print("Accuracy: {:.2%}".format(accuracy_percent))
print("Precision: {:.2%}".format(precision_percent))
print("Recall: {:.2%}".format(recall_percent))
print("F1 Score: {:.2%}".format(f1_percent))
```

## Lampiran 7 – Daftar Bimbingan

Bimbingan					
No	Dosen	Topik	Tanggal Bimbingan	Jenis Bimbingan	Catatan Perbaikan
1	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 6 desember bimbingan mengenai pengajuan proposal skripsi dan membuat latar belakang penelitian	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
2	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 27 Januari bimbingan mengeai proposal skripsi dari bab 1-3	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
3	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 13 Maret 2023 bimbingan mengenai tugas akhir dan bertanya mengenai bab 4	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
4	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada Tanggal 7 April 2023 Melakukan bimbingan bab 4	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
5	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 10 Juni 2023 melakukan bimbingan dan revisi bab 4	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
6	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 5 Juli 2023 bimbingan mengenai bab 4 dan memperbaiki pelabelan data	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
7	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 9 Juli 2023 bimbingan mengenai seluruh bab pada tugas akhir dan merevisi pelabelan data dan kesimpulan, saran	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
8	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 24 Juli bimbingan bab 4	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
9	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 26 Juli 2023 bimbingan revisi pada bab 4 dan 5.	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	
10	5709 - MUNAWAR , S.TP, MM, Ph.D.	Pada tanggal 27 Juli menyerahkan Tugas Akhir kepada pembimbing untuk di baca oleh pembimbing	30 Jul 2023	Skripsi/Tesis/BusinessPlan Proposal	