

PEMBOBOTAN KATA MENGGUNAKAN TF-IDF

Chapter 4

Sains Data



TF-IDF

Term Frequency-Inverse Document Frequency

- Metode pembobotan kata dalam dokumen yang digunakan dalam Natural Language Processing (NLP) dan Information Retrieval (IR).
- Teknik ini bertujuan untuk menilai sebuah kata dalam dokumen yang relatif terhadap kumpulan dokumen (corpus).

FORMULA TF

 Mengukur seberapa sering sebuah kata muncul dalam satu dokumen

$$TF(t,d) = rac{f(t,d)}{\sum_{t' \in d} f(t',d)}$$

- ullet = kata yang sedang dihitung frekuensinya
- d = dokumen tertentu
- ullet f(t,d) = jumlah kemunculan kata t dalam dokumen d
- $\sum_{t' \in d} f(t',d)$ = total jumlah kata dalam dokumen d

FORMULA IDF

 Mengukur seberapa jarang atau unik sebuah kata dalam seluruh koleksi dokumen. Kata yang muncul di banyak dokumen memiliki nilai IDF yang rendah, sementara kata yang jarang muncul memiliki nilai IDF yang tinggi

$$IDF(t,D) = \log rac{|D|}{1+|\{d\in D: t\in d\}|}$$

- D = jumlah total dokumen dalam kumpulan
- $ullet \ |\{d\in D:t\in d\}|$ = jumlah dokumen yang mengandung kata t
- Ditambahkan 1 dalam penyebut untuk menghindari pembagian dengan nol jika kata tidak ada di dokumen mana pun.

FORMULA TF-IDF

$$TF - IDF(t, d, D) = TF(t, d) \times IDF(t, D)$$

GaCL M=nM 10=1(x2-X)+(2-4)

JENIS TF-IDF

- TF-IDF Unigram \rightarrow ngram_range = (1,1)
- TF-IDF Unigram dan Bigram →
 ngram_range = (1,2)
- TF-IDF Bigram \rightarrow ngram_range = (2,2)
- TF-IDF Trigram \rightarrow ngram_range = (3,3)

LIB YANG DIGUNAKAN

Perhitungan TF-IDF pada Python menggunakan lib

Scikit-learn: "sklearn"

Codingnya

IMPORT DATA

```
import pandas as pd
import numpy as np

dm = pd.read_csv("tomlembong50_clean.csv", usecols=["tweet_clean", "Label"])
dm.columns = ["tweet_clean", "Label"]
dm.head(10)
```

MENJUMLAHKAN FITUR DAN STRING LIST

```
#checking the count of the dependent variable
dm['Label'].value_counts()
Positif
           26
Negatif
           24
Name: Label, dtype: int64
# convert list formated string to list
import ast
def convert text list(texts):
    texts = ast.literal_eval(texts)
    return [text for text in texts]
dm["tekslist"] = dm["tweet_clean"].apply(convert_text_list)
print(dm["tekslist"][23])
print("\ntype : ", type(dm["tekslist"][23]))
```

PERHITUNGAN TF

```
def calc TF(document):
   #perhitungan jumlah kata
   TF dict = {}
   for term in document:
       if term in TF dict:
           TF dict[term] += 1
       else:
           TF dict[term] = 1
   #perhitungan tf
   for term in TF dict:
       TF dict[term] = TF dict[term] / len(document)
   return TF dict
dm["TF dict"] = dm['tekslist'].apply(calc TF)
dm["TF dict"].head()
    {'acara': 0.14285714285714285, 'tahan': 0.1428...
  {'acara': 0.04, 'tom': 0.04, 'lembong': 0.04, ...
    {'acara': 0.03125, 'tom': 0.0625, 'lembong': 0...
    {'adil': 0.14285714285714285, 'hukum': 0.07142...
Name: TF dict, dtype: object
index = 23
print('%20s' % "term", "\t", "TF\n")
for key in dm["TF dict"][index]:
   print('%20s' % key, "\t", dm["TF dict"][index][key])
```

PERHITUNGAN IDF

```
#menghitung idf

n_document = len(dm)
def calc_IDF(__n_document, __DF):
    IDF_Dict = {}
    for term in __DF:
        IDF_Dict[term] = np.log(__n_document / (__DF[term] + 1))
    return IDF_Dict

#penyimpanan kamus idf
IDF = calc_IDF(n_document, DF)
```

```
#perhitungan TF-IDF
def calc_TF_IDF(TF):
    TF_IDF_Dict = {}
    for key in TF:
        TF_IDF_Dict[key] = TF[key] * IDF[key]
    return TF_IDF_Dict

#penyimpanan variabel TF-IDF
dm["TF-IDF_dict"] = dm["TF_dict"].apply(calc_TF_IDF)
```

HASIL PERHITUNGAN TFIDF

```
# memunculkan nilai TF-IDF
index = 23
print('%20s' % "term", "\t", '%10s' % "TF", "\t", '%20s' % "TF-IDF\n")
for key in dm["TF-IDF dict"][index]:
    print('%20s' % key, "\t", dm["TF_dict"][index][key] ,"\t" , dm["TF-IDF_dict"][index][key])
                term
                                 TF
                                                       TF-IDF
                azab
                         0.043478260869565216
                                                  0.13995112282035654
                 lho
                         0.043478260869565216
                                                  0.13995112282035654
                 mul
                         0.043478260869565216
                                                  0.13995112282035654
               saksi
                         0.043478260869565216
                                                  0.10981428888296763
                kali
                         0.043478260869565216
                                                  0.13995112282035654
               hidup
                                                  0.13995112282035654
                         0.043478260869565216
                kalo
                         0.043478260869565216
                                                  0.13995112282035654
                         0.043478260869565216
                  ga
                                                  0.10981428888296763
                         0.043478260869565216
                                                  0.12232220507652332
                  ya
                anak
                         0.043478260869565216
                                                  0.12232220507652332
               turun
                         0.043478260869565216
                                                  0.13995112282035654
                         0.043478260869565216
                                                  0.13995112282035654
                  mu
              nandur
                         0.043478260869565216
                                                  0.13995112282035654
             ngunduh
                         0.043478260869565216
                                                  0.13995112282035654
               mull1
                         0.043478260869565216
                                                  0.13995112282035654
               gusti
                         0.043478260869565216
                                                  0.13995112282035654
              mboten
                         0.043478260869565216
                                                  0.13995112282035654
                         0.043478260869565216
                                                  0.13995112282035654
                sare
                         0.043478260869565216
                                                  0.06583164054912068
               sabar
                         0.043478260869565216
                                                  0.13995112282035654
                 tom
                         0.043478260869565216
                                                  0.0036252873451761303
             lembong
                         0.043478260869565216
                                                  0.0026902349442646718
            keluarga
                         0.043478260869565216
                                                  0.13995112282035654
```

MATRIK TF-IDF

```
#matrik tf-idf
# pengurutan descending berdasarkan nilai DF
sorted DF = sorted(DF.items(), key=lambda kv: kv[1], reverse=True)[:30]
# pembuatan list kata dari pengurutan `sorted DF`
unique term = [item[0] for item in sorted DF]
def calc TF IDF Vec( TF IDF Dict):
   TF IDF vector = [0.0] * len(unique term)
   # For each unique word, if it is in the review, store its TF-IDF value.
   for i, term in enumerate(unique term):
       if term in   TF IDF Dict:
           TF_IDF_vector[i] = __TF_IDF_Dict[term]
   return TF IDF vector
dm["TF_IDF_Vec"] = dm["TF-IDF_dict"].apply(calc_TF_IDF_Vec)
print("tampil baris pertama matrix TF IDF Vec Series\n")
print(dm["TF IDF Vec"][0])
print("\nukuran matrix : ", len(dm["TF IDF Vec"][0]))
```

MENAMPILKAN KI DALAM LIST

```
#menampilkan top 30 term tf-idf

# konversi ke dalam List

TF_IDF_Vec_List = np.array(dm["TF_IDF_Vec"].to_list())

sums = TF_IDF_Vec_List.sum(axis=0)

data = []

for col, term in enumerate(unique_term):
    data.append((term, sums[col]))

ranking = pd.DataFrame(data, columns=['term', 'rank'])
ranking.sort_values('rank', ascending=False)
```

rank

3		2.154247
2	anies	1.466529
26	aja	1.451356
14	bebas	1.404827
8	praperadilan	1.331056
5	gula	1.238108
7	impor	1.169731
6	sangka	1.017534
4	уд	1.015680
23	adil	0.959909
13	bukti	0.947504
10	jagung	0.927870
9	hukum	0 918525

term











NEXT.... PEMODELAN DENGAN ALGORITMA MACHINE LEARNING