

*Muhammad Adam Zaky Jiddyansah*

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# ***Pencarian dan Penambahan Web***

To blah, blah, and blah.

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## *Selamat Datang di Jupyter Book Saya*

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Selamat Datang di Quarto Book Saya Website ini menyajikan beragam konten terkait Penambangan Data atau Data Mining. Mulai dari tugas, materi, hingga artikel-artikel menarik yang mencakup dua tema utama: Pencarian Penambangan informasi melalui Web dan teknik-teknik penambangan data yang relevan. Jelajahi menu di samping untuk menemukan lebih banyak wawasan dan praktik terkait, serta kunjungi Website Personal(<https://adamzakys.github.io/>) untuk mengetahui biodata saya secara lebih rinci. Selamat menikmati pembelajaran dan penjelajahan!



# 1

## *tokenisasi ulang*

##Import dan Install modul yang diperlukan

```
!pip install nltk
!pip install Sastrawi
!pip install gensim
import pandas as pd
import re
import nltk
import numpy as np
from gensim.models import Word2Vec
from gensim.models import LdaModel
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.tokenize import RegexpTokenizer
from Sastrawi.Stemmer.StemmerFactory import StemmerFactory
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer
from sklearn.preprocessing import OneHotEncoder
from gensim import corpora, models
```

Requirement already satisfied: nltk in /usr/local/lib/python3.10/dist-packages (3.8.1)

Requirement already satisfied: click in /usr/local/lib/python3.10/dist-packages (from nltk) (8.1.7)

Requirement already satisfied: joblib in /usr/local/lib/python3.10/dist-packages (from nltk) (1.3.2)

Requirement already satisfied: regex>=2021.8.3 in /usr/local/lib/python3.10/dist-packages (from nltk)

Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from nltk) (4.66.1)

Collecting Sastrawi

Downloading Sastrawi-1.0.1-py2.py3-none-any.whl (209 kB)

209.7/209.7 kB 2.9 MB/s eta 0:00:00

Installing collected packages: Sastrawi

Successfully installed Sastrawi-1.0.1

Requirement already satisfied: gensim in /usr/local/lib/python3.10/dist-packages (4.3.2)

Requirement already satisfied: numpy>=1.18.5 in /usr/local/lib/python3.10/dist-packages (from gensim)

Requirement already satisfied: scipy>=1.7.0 in /usr/local/lib/python3.10/dist-packages (from gensim)

Requirement already satisfied: smart-open>=1.8.1 in /usr/local/lib/python3.10/dist-packages (from gensim)

##Konfigurasi nltk

```
nltk.download("punkt")
nltk.download("stopwords")
```

```
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt.zip.
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data]   Unzipping corpora/stopwords.zip.
```

True

##mengambil data pta yang sudah dicrawl di program sebelumnya

```
from google.colab import drive
drive.mount('/content/drive')

file_path = "/content/drive/My Drive/Penambangan WEB/Data/full_210411100234_data_pgsc.csv"

df = pd.read_csv(file_path)
df
```

Mengubah teks menjadi lower

```
df=df.astype(str)
df["Abstrak"] = df["Abstrak"].apply(lambda x: x.lower())

abstrak_column = df["Abstrak"]
```

##Tokenisasi menggunakan word tokenize pemisahan kata dalam suatu kalimat dengan tujuan untuk proses analisis teks lebih lanjut

```
def process_tokenize(text):
    text = text.split()
    return text

processed_abstrak = abstrak_column.apply(process_tokenize)

df["processed_abstrak"] = processed_abstrak

data = pd.DataFrame(df, columns=['processed_abstrak'])
data
```

##Punctuation Menghilangkan tanda baca



```
def process_punctuation(tokens):
    cleaned_tokens = [re.sub(r'[.,():-]', '', token) for token in tokens]
    cleaned_tokens = [re.sub(r'\d+', '', token) for token in cleaned_tokens]

    return cleaned_tokens

df['processed_abstrak'] = df['processed_abstrak'].apply(process_punctuation)

data = pd.DataFrame(df, columns=['processed_abstrak'])
data
```

##mensaring kata2 dengan Stopword

```
def process_stopword_token(tokens):
    stop_words = set(stopwords.words("indonesian"))
    filtered_tokens = [token for token in tokens if token.lower() not in stop_words]
    return " ".join(filtered_tokens)

df['processed_abstrak'] = df['processed_abstrak'].apply(process_stopword_token)

data = pd.DataFrame(df, columns=['processed_abstrak'])
```

##Proses menghilangkan imbuhan baik yang berada di awal kata ataupun yang berada di akhir(Steeming) contoh : kata “running” dan “runner” dapat di-stem menjadi “run”.

##Feature Extraction melakukan perhitungan dan perbandingan yang bisa digunakan untuk mengklasifikasikan ciri-ciri yang dimiliki

mengonversi koleksi dokumen teks menjadi representasi numerik

```
countvectorizer = CountVectorizer(analyzer= 'word', stop_words='english')
count_wm = countvectorizer.fit_transform(df['processed_abstrak'])
count_tokens = countvectorizer.get_feature_names_out()
df_countvect = pd.DataFrame(data = count_wm.toarray(),columns = count_tokens)
print('Count Vectorizer\n')
df_countvect
```

mengonversi teks menjadi representasi numerik yang lebih informatif

##SKIP GRAM Word2Vec

```
#Menambahkan kode untuk melatih model Word2Vec (Skip-gram)
```

```
sentences = df['processed_abstrak'].apply(lambda x: x.split()).tolist()
model = Word2Vec(sentences, vector_size=100, window=5, min_count=1, sg=1)
```

```
#contoh penggunaan skip gram pada dataset abstrak yang sudah dinormalisasi
print("Model Word2Vec (Skip-gram) berhasil dilatih.")
print("Contoh penggunaan:")
word_vector = model.wv['akurat']
print("Vektor kata 'kata_contoh':", word_vector)
similar_words = model.wv.most_similar('akurat')
print("Kata-kata yang mirip dengan 'akurat':", similar_words)
```

### ##LDA

```
documents = df['processed_abstrak'].apply(lambda x: x.split())

# Membuat representasi teks dalam bentuk "bag of words" / konsep yang diambil dari analisis te
dictionary = corpora.Dictionary(documents)
corpus = [dictionary.doc2bow(doc) for doc in documents]

# create and training model LDA
lda_model = LdaModel(corpus, num_topics=3, id2word=dictionary, passes=15)

# Menghitung proporsi kata dalam setiap topik
topic_word_proposals = lda_model.get_topics()

# Membuat DataFrame untuk menyimpan proporsi kata dalam topik
topic_word_proposals_df = pd.DataFrame(topic_word_proposals, columns=[dictionary[i] for i in range(l

# Menyimpan DataFrame proporsi kata dalam topik ke dalam file CSV
topic_word_proposals_df.to_csv("proporsi_kata_dalam_topik.csv", index=False)

# Menghitung proporsi topik dalam dokumen
document_topic_proposals = [lda_model.get_document_topics(doc) for doc in corpus]

# Membuat DataFrame untuk menyimpan proporsi topik dalam dokumen
document_topic_proposals_df = pd.DataFrame(columns=["Judul"] + [f"Topic {i+1}" for i in range(1

for i, doc_topic_proposals in enumerate(document_topic_proposals):
    row_data = {"Judul": df['Judul'].iloc[i]} # Menggunakan kolom "Judul" dari DataFrame awal
    for topic, prop in doc_topic_proposals:
        row_data[f"Topic {topic + 1}"] = prop
    # Menambahkan baris data ke DataFrame
    document_topic_proposals_df = pd.concat([document_topic_proposals_df, pd.DataFrame([row_data,
```

```
# ubah nilai NaN dengan 0
document_topic_proposals_df = document_topic_proposals_df.fillna(0)

document_topic_proposals_df
```

### ##TF-IDF

```
# Melakukan pra-pemrosesan teks untuk TF-IDF
documents_tfidf = df['processed_abstrak'] # Menggunakan processed_abstrak2 yang sudah diproses

# Melatih model TF-IDF Vectorizer
tfidfvectorizer = TfidfVectorizer(analyzer='word', stop_words='english')
tfidf_wm = tfidfvectorizer.fit_transform(documents_tfidf)
tfidf_tokens = tfidfvectorizer.get_feature_names_out()

# Mengubah matriks TF-IDF menjadi DataFrame
df_tfidfvect = pd.DataFrame(data=tfidf_wm.toarray(), columns=tfidf_tokens)

# Menambahkan kolom dokumen ke DataFrame TF-IDF
df_tfidfvect.insert(0, 'Judul', df['Judul']) # Menggunakan kolom 'Judul' sebagai dokumen

print('TF-IDF Vectorizer\n')
df_tfidfvect
```

### ##Perbandingan silhouette score antara LDA dan TF-IDF setelah Cluster

```
from sklearn.metrics import silhouette_score

# Jumlah kluster yang diinginkan
n_clusters = 3

# Melakukan clustering pada data hasil LDA
kmeans_lda = KMeans(n_clusters=n_clusters, random_state=0, n_init=10)
kmeans_lda.fit(document_topic_proposals_df.iloc[:, 1:]) # Menggunakan proporsi topik sebagai f

# Melakukan clustering pada data TF-IDF
kmeans_tfidf = KMeans(n_clusters=n_clusters, random_state=0, n_init=10)
kmeans_tfidf.fit(tfidf_wm)

# Menghitung Silhouette Score untuk LDA
silhouette_lda = silhouette_score(document_topic_proposals_df.iloc[:, 1:], kmeans_lda.labels_)

# Menghitung Silhouette Score untuk TF-IDF
```

```

silhouette_tfidf = silhouette_score(tfidf_wm, kmeans_tfidf.labels_)

# Menambahkan label kluster ke DataFrame hasil LDA
document_topic_proposals_df['Cluster_LDA'] = kmeans_lda.labels_

# Menambahkan label kluster ke DataFrame hasil TF-IDF
df_tfidfvect['Cluster_TFIDF'] = kmeans_tfidf.labels_

# Menampilkan hasil clustering LDA dan TF-IDF
print("Hasil Clustering LDA:")
print(document_topic_proposals_df[['Judul', 'Cluster_LDA']])
print("\nHasil Clustering TF-IDF:")
print(df_tfidfvect[['Judul', 'Cluster_TFIDF']])
print("=====")
# Menampilkan hasil perbandingan
print(f"Silhouette Score untuk LDA: {silhouette_lda}")
print(f"Silhouette Score untuk TF-IDF: {silhouette_tfidf}")

```

#### #Simpan Hasil Preprocessing Data

```

import pandas as pd

svdf = pd.DataFrame(df)
svdf.to_csv("210411100234_data_pgsd_with_preprocessing.csv", index=False)
document_topic_proposals_df.to_csv("210411100234_ModelingTopic.csv", index=False)

from google.colab import drive
import shutil

drive.mount('/content/drive')

source_path = "210411100234_data_pgsd_with_preprocessing.csv"
destination_path = "/content/drive/My Drive/Penambangan WEB/Data/210411100234_data_pgsd_with_preprocessing.csv"

source_path = "210411100234_ModelingTopic.csv"
destination_path = "/content/drive/My Drive/Penambangan WEB/Data/210411100234_ModelingTopic.csv"

shutil.copy(source_path, destination_path)

```

## 2

---

### *Crawl Berita Kompas*

---

Impor Modul:

requests: Digunakan untuk membuat permintaan HTTP ke situs web. BeautifulSoup dari bs4: Sebuah pustaka untuk mengekstrak data dari file HTML dan XML.

pandas: Sebuah pustaka untuk manipulasi data dalam analisis data.

Inisialisasi:

Skrip menginisialisasi kamus kosong bernama csv dengan kunci “Judul”, “Berita”, dan “Kategori” untuk menyimpan data yang diambil. Daftar categories berisi kategori-kategori untuk artikel berita yang akan diambil. Loop Pencarian Berita:

Skrip melakukan iterasi melalui setiap kategori dan nomor halaman untuk membentuk URL halaman indeks berita di situs Kompas. Kemudian, skrip mengirim permintaan HTTP ke URL yang dibentuk dan menguraikan konten HTML menggunakan BeautifulSoup.

Ekstraksi Artikel:

Untuk setiap halaman, skrip mencari semua elemen div dengan kelas “latest-indeks mt2 clearfix” yang berisi informasi tentang artikel. Selanjutnya, skrip mengekstrak tautan artikel dan mengambil konten masing-masing artikel dengan mengirim permintaan HTTP lainnya.

Ekstraksi Data:

Skrip mengekstrak judul dan isi dari setiap artikel menggunakan BeautifulSoup berdasarkan struktur HTML situs web Kompas. Data yang diekstrak kemudian ditambahkan ke daftar yang sesuai dalam kamus csv. Membuat DataFrame:

Setelah mengambil semua artikel, skrip mengonversi data yang dikumpulkan ke dalam DataFrame pandas yang dinamai data.

Menyimpan ke CSV:

DataFrame kemudian disimpan ke file CSV bernama

“Data\_Berita\_All\_Kategori.csv” pada path yang ditentukan (“/content/drive/My Drive/Penambangan WEB/Data/”).

```
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive")

```
import requests
from bs4 import BeautifulSoup as soup
import pandas as pd

csv = {"Judul": [], "Berita": [], "Kategori": []}

categories = ["bola", "lifestyle", "umkm", "tekno"]

for kategori in categories:
    for i in range(1, 100):
        url = f"https://indeks.kompas.com/?site={kategori}&page={i}"
        print(f"Crawling: {url}")
        client = requests.get(url)
        page_html = client.content
        page_soup = soup(page_html, "html.parser")
        berita = page_soup.findAll("div", {"class": "latest--indeks mt2 clearfix"})

        print(f"Found {len(berita)} articles on this page.")

    for h in berita:
        link_berita = h.select_one('a.article__link')
        if link_berita:
            link_berita = link_berita['href']
            r = requests.get(link_berita)
            page = soup(r.content, "html.parser")
            halaman_isi = page.select_one("div", {"class": "col-bs10-10"})

            judul_berita = halaman_isi.select("h1", {"class": "read__title"})
            judul = judul_berita[0].text

            isi_berita = halaman_isi.findAll('p')
            isi = '\n'.join([p.get_text() for p in isi_berita])

            csv["Judul"].append(judul)
            csv["Berita"].append(isi)
            csv["Kategori"].append(kategori)
```

```
data = pd.DataFrame(csv)
csv_path_drive = '/content/drive/My Drive/Penambangan WEB/Data/Data_Berita_All_Kategori.csv'
data.to_csv(csv_path_drive, index=False)
```

```
Crawling: https://indeks.kompas.com/?site=bola&page=1
Found 1 articles on this page.
Crawling: https://indeks.kompas.com/?site=bola&page=2
Found 1 articles on this page.
Crawling: https://indeks.kompas.com/?site=bola&page=3
Found 1 articles on this page.
Crawling: https://indeks.kompas.com/?site=bola&page=4
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```

```

drive.mount('/content/drive')
file_path = "/content/drive/My Drive/Penambangan WEB/Data/Data_Berita_All_Kategori.csv"
df = pd.read_csv(file_path)
df

```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive")

|   | Judul   | Berita                                    |
|---|---|---|
| 0 | Hasil Persib Vs PSM Makassar 0-0: Gagal Menang... | Hasil Persib Vs PSM Makassar 0-0: Gagal M |
| 1 | Manchester City Pantau Bintang Argentina yang ... | Manchester City Pantau Bintang Argentina  |
| 2 | Sepak Terjang Paris Brunner, Sang Pemain Terba... | Sepak Terjang Paris Brunner, Sang Pemain  |
| 3 | Musim Gemilang Astra Honda Racing Team, Domina... | Musim Gemilang Astra Honda Racing Team    |



|     | Judul   | Berita                                    |
|-----|---|---|
| 4   | Pembelajaran dari Piala Dunia U17 2023 demi Se... | Pembelajaran dari Piala Dunia U17 2023 de |
| ... | ...   | ...                                       |
| 391 | Fitur di Smartwatch Huawei Watch GT 4 Bisa Pan... | Fitur di Smartwatch Huawei Watch GT 4 B   |
| 392 | Hands-on Gelang Pintar Xiaomi Smart Band 8, Ga... | Hands-on Gelang Pintar Xiaomi Smart Ban   |
| 393 | Google Rayakan Ulang Tahun Ke-25, Ini Sejarah ... | Google Rayakan Ulang Tahun Ke-25, Ini Se  |
| 394 | Social Commerce dan E-commerce, Apa Bedanya? B... | Social Commerce dan E-commerce, Apa Bec   |
| 395 | Samsung Sudah Siapkan Galaxy S25, Ini Buktinya    | Samsung Sudah Siapkan Galaxy S25, Ini Bu  |



# 3

## *Tugas Crawling\_berita\_Kompas dan ekstrak kata kunci*

- Nama : Muhammad Adam Zaky Jiddyansah
- NIM : 210411100234
- Kelas : Penambangan dan Pencarian Web B

### 3.1 Menginstall dan mengimport library yang dibutuhkan

Libraries yang Digunakan \* BeautifulSoup (bs4): Library untuk melakukan web scraping dan ekstraksi data HTML. \* Requests: Library untuk membuat permintaan HTTP ke situs web dan mendapatkan konten halaman. \* Pandas (pd): Library untuk mengelola dan menyimpan data dalam format CSV.

```
from bs4 import BeautifulSoup as soup
import requests
import pandas as pd
import nltk
from nltk.tokenize import sent_tokenize
import re
from nltk.corpus import stopwords
from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer, TfidfTransformer
from sklearn.metrics.pairwise import cosine_similarity
from nltk.tokenize import word_tokenize
from collections import Counter
import networkx as nx
import matplotlib.pyplot as plt

nltk.download("punkt")
nltk.download("stopwords")
```

[nltk\_data] Downloading package punkt to /root/nltk\_data...

```
[nltk_data] Unzipping tokenizers/punkt.zip.
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.

True
```

### 3.2 Web Scraping Berita Kompas dan Penyimpanan ke Google Drive

Kode ini menggunakan teknik web scraping untuk mengumpulkan judul dan isi berita dari 200 halaman indeks situs Kompas.com. Berikut adalah penjelasan komponen utama dari kode tersebut:

**Proses Web Scraping dan menyimpan data** \* Iterasi sebanyak 5 (karna yang akan dipreprocessing kali ini cuman 1 berita) halaman indeks(tergantung indeks pada website) di situs Kompas.com. \* Untuk setiap halaman, melakukan permintaan HTTP dan mendapatkan konten HTML. \* Menggunakan BeautifulSoup untuk mengekstrak daftar berita dari halaman tersebut. \* Untuk setiap berita, mengakses halaman individu untuk mengambil judul dan isi berita. \* Data judul dan isi berita ditambahkan ke dalam dictionary csv. Penyimpanan Data \* Setelah proses scraping selesai, data disimpan dalam file CSV dengan nama "Data\_BeritaKompas.csv" terlebih dahulu di environment Colab. \* File CSV tersebut disalin ke dalam Google Drive pada path "/content/drive/My Drive/Penambangan WEB/Data/".

```
from google.colab import drive
drive.mount('/content/drive')

csv = {"Judul": [], "Berita": []}

for i in range(1, 20):
    url = "https://indeks.kompas.com/?page={}".format(i)
    client = requests.get(url)
    page_html = client.content
    page_soup = soup(page_html, "html.parser")
    berita = page_soup.findAll("div", {"class": "latest--indeks mt2 clearfix"})

    for h in berita:
        r = requests.get(h.select_one('a.article__link')['href'])
        page = soup(r.content, "html.parser")
        halaman_isi = page.select_one("div", {"class": "col-bs10-10"})
```

```

judul_berita = halaman_isi.select("h1", {"class": "read__title"})
judul = judul_berita[0].text

isi_berita = halaman_isi.findAll('p')
isi = '\n'.join([p.get_text() for p in isi_berita])

csv["Judul"].append(judul)
csv["Berita"].append(isi)

data = pd.DataFrame(csv)
csv_path_drive = '/content/drive/My Drive/Penambangan WEB/Data/Data_BeritaKompas.csv'
data.to_csv(csv_path_drive, index=False)

```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive")

### 3.3 Import dan Membaca Data CSV

Menggunakan Pandas untuk membaca file CSV yang telah disimpan sebelumnya. File CSV tersebut berisi data judul dan isi berita dari Kompas.com yang telah di-web scrape sebelumnya, lalu menampilkan data yang telah dibaca dari file CSV dalam bentuk DataFrame untuk dapat dianalisis lebih lanjut.

```

from google.colab import drive
drive.mount('/content/drive')
file_path = "/content/drive/My Drive/Penambangan WEB/Data/Data_BeritaKompas.csv"
df = pd.read_csv(file_path)
df

```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive")

|   | Judul   | Berita                                   |
|---|---|--|
| 0 | Lirik dan Makna Lagu Oh Indang Oh Apang, Lagu ... | Lirik dan Makna Lagu Oh Indang Oh Apa    |
| 1 | Berapa Banyak Kalori yang Dibakar dengan Berja... | Berapa Banyak Kalori yang Dibakar deng   |
| 2 | Pendaftaran Petugas KPPS Pemilu 2024 Dibuka, B... | Pendaftaran Petugas KPPS Pemilu 2024 I   |
| 3 | Produser Ungkap Tiket Advance Siksa Neraka Sol... | Produser Ungkap Tiket Advance Siksa Ne   |
| 4 | Peringati Hari Korban 40 Ribu Jiwa, Pj Gubernu... | Peringati Hari Korban 40 Ribu Jiwa, Pj G |
| 5 | Lirik dan Chord Lagu Salam Kenal - Vidi Aldiano   | Lirik dan Chord Lagu Salam Kenal - Vidi  |
| 6 | Pikap Tabrak Truk Tronton di Kota Malang, Satu... | Pikap Tabrak Truk Tronton di Kota Mala   |
| 7 | Perbandingan Yamaha XMAX 250 Tech Max dan Kymc... | Perbandingan Yamaha XMAX 250 Tech M      |
| 8 | Hasil Mediasi Keluarga Bayi HNM: RS Hermina Po... | Hasil Mediasi Keluarga Bayi HNM: RS He   |
| 9 | Tanggapi Hasil Survei Litbang "Kompas", Mahfud... | Tanggapi Hasil Survei Litbang "Kompas",  |

|    | Judul   | Berita                                   |
|----|---|--|
| 10 | Kepsek SMA 17 Makassar Dimutasi Dampak Aksi De... | Kepsek SMA 17 Makassar Dimutasi Damp     |
| 11 | Status WhatsApp Bakal Bisa Dibagikan Langsung ... | Status WhatsApp Bakal Bisa Dibagikan L   |
| 12 | Alat yang Bekerja Berdasarkan Prinsip Tuas        | Alat yang Bekerja Berdasarkan Prinsip Tu |
| 13 | Pelaku Perusakan Belasan Mobil Dinas di Semara... | Pelaku Perusakan Belasan Mobil Dinas di  |
| 14 | Rafael Alun Dihukum Bayar Uang Pengganti 18,9 ... | Rafael Alun Dihukum Bayar Uang Pengga    |
| 15 | Begini Peran Teknik Industri dalam Pemanfaatan... | Begini Peran Teknik Industri dalam Pema  |
| 16 | Lexus Siapkan RZ550e, RZ Berperforma Tinggi       | Lexus Siapkan RZ550e, RZ Berperforma T   |
| 17 | Pengendara Motor Tewas di Lokasi Kejadian Sete... | Pengendara Motor Tewas di Lokasi Kejad   |
| 18 | Newcastle Vs Milan, Kabar Baik untuk Rossoneri... | Newcastle Vs Milan, Kabar Baik untuk Ro  |

### 3.4 Mengambil salah satu berita untuk menjadi sampel data yang akan dianalisis

```
berita = df['Berita'].iloc[15]
berita
```

'Begini Peran Teknik Industri dalam Pemanfaatan AI di Era Industri 4.0\n\nKOMPAS.com - Implementasi kecerdasan buatan atau artificial intelligence (AI) dalam industri menjadi salah satu poi

### 3.5 Pra-pemrosesan Teks / Preprocessing

Mengambil teks sebagai input dan melakukan beberapa langkah pra-pemrosesan pada teks dalam bahasa Indonesia. Langkah-langkah tersebut mencakup penghapusan angka, pembersihan karakter khusus, konversi teks ke huruf kecil, penghapusan kata-kata umum (stopwords), dan penggabungan kata-kata yang telah melalui proses sebelumnya. Tujuannya adalah membersihkan dan mempersiapkan teks agar dapat digunakan lebih lanjut dalam analisis atau pemrosesan lanjutan.

```
def preprocessing(text):
    text = re.sub(r'\d+', '', text)
    text = re.sub(r'^\w\s.', '', text)
    text = text.lower()

    stop_words = set(stopwords.words('indonesian'))
```

```

words = text.split()
filtered_words = [word for word in words if word.lower() not in stop_words]

preprocessing_text = ' '.join(filtered_words)

return preprocessing_text

berita = preprocessing(berita)
print(berita)

```

peran teknik industri pemanfaatan ai era industri . kompas.com implementasi kecerdasan buatan artific

##Memisahkan kalimat dengan word tokenize

Tokenisasi kalimat adalah proses memecah sebuah teks menjadi kalimat-kalimat yang lebih kecil atau unit yang disebut token, di mana token dapat berupa kata, frasa, atau karakter. Word tokenization, atau tokenisasi kata, adalah jenis tokenisasi yang fokus pada memecah teks menjadi unit kata.

```

kalimat = nltk.sent_tokenize(berita)
kalimat = [sentence.replace('.', ' ') for sentence in kalimat]
print(kalimat)

```

['peran teknik industri pemanfaatan ai era industri ', 'kompascom implementasi kecerdasan buatan artif

```

kata = word_tokenize(berita)
kata = [k.lower() for k in kata if k != '.']
kata = list(set(kata))
print(kata)

```

['dihadapi', 'as', 'merangkul', 'pemborosan', 'printing', 'solusi', 'perguruan', 'signifikan', 'konsu

```

# Inisialisasi DataFrame
matriks_kata = pd.DataFrame(0, index=kata, columns=kata)

```

```

for sentence in kalimat:
    words = sentence.split()
    for i in range(len(words)-1):
        current_word = words[i]
        next_word = words[i+1]

# Pastikan kedua kata ada dalam indeks matrikskata sebelum mengupdate nilai

```

```
if current_word in matriks_kata.index and next_word in matriks_kata.columns:
    matriks_kata.at[current_word, next_word] += 1
```

```
matriks_kata
```

|             | dihadapi | as  | merangkul | pemborosan | printing | solusi | perguruan | signifikan | konsums |
|-------------|----------|-----|-----------|------------|----------|--------|-----------|------------|---------|
| dihadapi    | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| as          | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| merangkul   | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| pemborosan  | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| printing    | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| ...         | ...      | ... | ...       | ...        | ...      | ...    | ...       | ...        | ...     |
| operasional | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| industrial  | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| panji       | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| menekankan  | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |
| computer    | 0        | 0   | 0         | 0          | 0        | 0      | 0         | 0          | 0       |

### 3.6 Perhitungan Kemiripan Kosinus

Menghitung kemiripan kosinus antara semua pasangan kalimat menggunakan matriks TF-IDF yang telah dihasilkan sebelumnya. Kemudian, hasil kemiripan kosinus tersebut disimpan dalam DataFrame untuk kemudahan analisis.

**Rumus Perhitungan Kemiripan Kosinus:** \* Untuk dua vektor A dan B, kemiripan kosinus dihitung dengan rumus

$$\text{Similarity}(A, B) = \frac{A \cdot B}{\|A\| \cdot \|B\|}$$

- Di sini,  $\cdot$  adalah operasi perkalian dot (dot product), dan  $\|A\|$  adalah norma Euclidean. Dalam konteks ini, vektor A dan B adalah vektor representasi TF-IDF dari dua kalimat. Hasilnya adalah skalar yang menunjukkan seberapa mirip dua kalimat tersebut, dengan nilai 1 menunjukkan kemiripan sempurna.

```
cosine = cosine_similarity(matriks_kata, matriks_kata)
```

```
similarity = pd.DataFrame(cosine, columns=matriks_kata.index, index=matriks_kata.index)
similarity
```



### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity35

|             | dihadapi | as  | merangkul | pemborosan | printing | solusi | perguruan | signifikan | konsum |
|-------------|----------|-----|-----------|------------|----------|--------|-----------|------------|--------|
| dihadapi    | 1.0      | 0.0 | 1.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| as          | 0.0      | 1.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| merangkul   | 1.0      | 0.0 | 1.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| pemborosan  | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| printing    | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| ...         | ...      | ... | ...       | ...        | ...      | ...    | ...       | ...        | ...    |
| operasional | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| industrial  | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| panji       | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| menekankan  | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |
| computer    | 0.0      | 0.0 | 0.0       | 0.0        | 0.0      | 0.0    | 0.0       | 0.0        | 0.0    |

Warning: Total number of columns (301) exceeds max\_columns (20) limiting to first (20) columns.

### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity

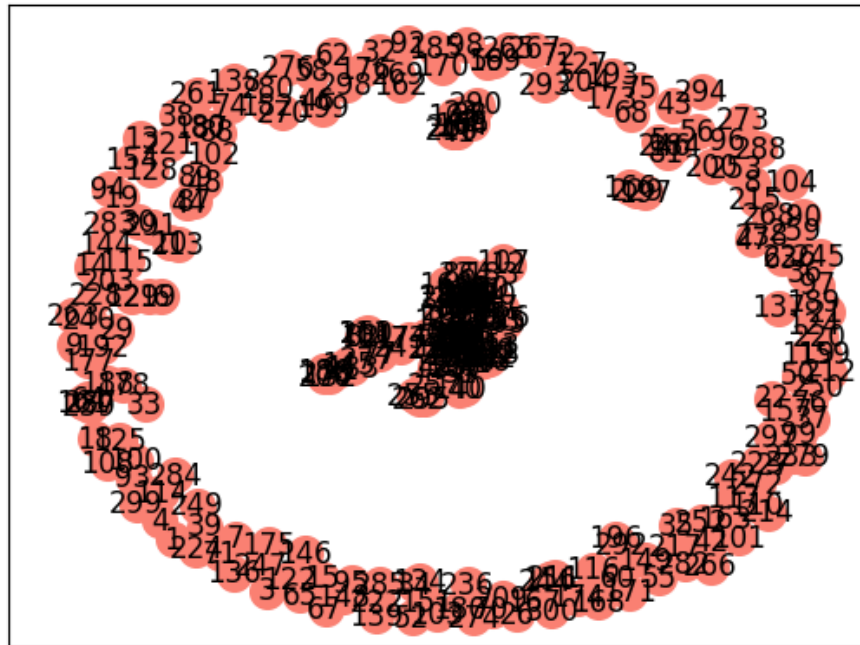
Membuat graf yang merepresentasikan hubungan antara kalimat-kalimat berdasarkan nilai cosine similarity. Setiap kalimat direpresentasikan sebagai node, dan garis yang menghubungkan dua node menunjukkan tingkat kemiripan antara kalimat tersebut. Semakin panjang dan tebal garis, semakin tinggi tingkat kemiripan antara dua kalimat. Graf ini membantu memvisualisasikan struktur hubungan antar kalimat dalam teks.

```
G = nx.DiGraph()
for i in range(len(cosine)):
    G.add_node(i)

for i in range(len(cosine)):
    for j in range(len(cosine)):
        similarity = cosine[i][j]
        if similarity > 0.1 and i != j:
            G.add_edge(i, j)

pos = nx.spring_layout(G)
nx.draw_networkx_nodes(G, pos, node_size=200, node_color='salmon')
nx.draw_networkx_edges(G, pos, edge_color='red', arrows=True)
nx.draw_networkx_labels(G, pos)
```

```
plt.show()
```



#Menemukan Kata kunci

##Page Rank **PageRank** adalah algoritma yang dikembangkan oleh Larry Page dan Sergey Brin, pendiri Google. Algoritma ini digunakan oleh mesin pencari Google untuk menentukan peringkat atau relevansi halaman web dalam hasil pencarian. Ide utama di balik PageRank adalah bahwa halaman web yang banyak di-link oleh halaman web lain dianggap lebih penting dan memiliki peringkat yang lebih tinggi.

### Konsep Dasar:

Prinsip Dasar: PageRank bekerja berdasarkan prinsip bahwa sebuah tautan dari halaman A ke halaman B dapat dianggap sebagai suara atau dukungan dari halaman A untuk halaman B. Semakin banyak tautan yang menuju ke sebuah halaman, semakin besar “nilai suara” atau “nilai dukungan” yang diberikan oleh halaman tersebut.

Tautan Kembali (Backlink): Pentingnya suatu halaman tidak hanya bergantung pada jumlah tautan masuk, tetapi juga pada kualitas tautan tersebut. Tautan dari halaman dengan peringkat tinggi lebih berharga daripada tautan dari halaman dengan peringkat rendah. Rumus dasar PageRank dapat

### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity 37

direpresentasikan sebagai sistem persamaan linear. Misalkan ada N halaman web, dan  $PR(A)$  adalah PageRank halaman A. Persamaan PageRank untuk halaman A adalah :

$$[PR(A) = (1-d) + d \left( \frac{PR(B)L(B)}{L(B)} + \frac{PR(C)L(C)}{L(C)} + \dots + \frac{PR(N)L(N)}{L(N)} \right)]$$

Dengan :

(d) adalah faktor damping (biasanya diatur sekitar 0.85)."

( $PR(B)$ ,  $PR(C)$ ,  $\dots$ ,  $PR(N)$ ) adalah PageRank dari halaman-halaman yang terhubung ke halaman (A)."

( $L(B)$ ,  $L(C)$ ,  $\dots$ ,  $L(N)$ ) adalah jumlah tautan keluar dari halaman (B, C,  $\dots$ , N)."

```
pagerank = nx.pagerank(G)

sorted_pagerank= sorted(pagerank.items(), key=lambda x: x[1], reverse=True)
print("Page Rank :")
for node, pagerank in sorted_pagerank:
    print(f"Node {node}: {pagerank:.4f}")
```

Page Rank :

```
Node 194: 0.0188
Node 208: 0.0167
Node 251: 0.0163
Node 164: 0.0156
Node 231: 0.0129
Node 79: 0.0126
Node 69: 0.0126
Node 198: 0.0125
Node 232: 0.0112
Node 61: 0.0112
Node 235: 0.0106
Node 254: 0.0103
Node 11: 0.0096
Node 227: 0.0096
Node 0: 0.0091
Node 2: 0.0091
Node 16: 0.0091
Node 57: 0.0091
Node 141: 0.0091
Node 158: 0.0091
Node 210: 0.0091
Node 256: 0.0091
```

Node 281: 0.0091  
Node 289: 0.0091  
Node 91: 0.0090  
Node 20: 0.0089  
Node 195: 0.0089  
Node 246: 0.0084  
Node 25: 0.0084  
Node 80: 0.0084  
Node 229: 0.0084  
Node 243: 0.0084  
Node 24: 0.0076  
Node 54: 0.0076  
Node 83: 0.0076  
Node 109: 0.0076  
Node 113: 0.0076  
Node 150: 0.0076  
Node 179: 0.0076  
Node 275: 0.0076  
Node 147: 0.0073  
Node 31: 0.0067  
Node 277: 0.0067  
Node 41: 0.0064  
Node 216: 0.0064  
Node 264: 0.0064  
Node 241: 0.0059  
Node 6: 0.0057  
Node 10: 0.0057  
Node 30: 0.0057  
Node 34: 0.0057  
Node 44: 0.0057  
Node 46: 0.0057  
Node 47: 0.0057  
Node 48: 0.0057  
Node 59: 0.0057  
Node 63: 0.0057  
Node 64: 0.0057  
Node 87: 0.0057  
Node 89: 0.0057  
Node 134: 0.0057  
Node 161: 0.0057  
Node 162: 0.0057  
Node 169: 0.0057  
Node 196: 0.0057  
Node 199: 0.0057  
Node 211: 0.0057

### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity<sup>39</sup>

Node 213: 0.0057  
Node 219: 0.0057  
Node 226: 0.0057  
Node 238: 0.0057  
Node 239: 0.0057  
Node 244: 0.0057  
Node 269: 0.0057  
Node 291: 0.0057  
Node 292: 0.0057  
Node 81: 0.0056  
Node 184: 0.0056  
Node 181: 0.0055  
Node 28: 0.0055  
Node 132: 0.0054  
Node 186: 0.0054  
Node 206: 0.0054  
Node 278: 0.0054  
Node 82: 0.0052  
Node 111: 0.0052  
Node 131: 0.0052  
Node 260: 0.0052  
Node 207: 0.0045  
Node 183: 0.0045  
Node 88: 0.0044  
Node 166: 0.0044  
Node 187: 0.0044  
Node 197: 0.0044  
Node 133: 0.0043  
Node 145: 0.0043  
Node 155: 0.0043  
Node 258: 0.0043  
Node 53: 0.0042  
Node 126: 0.0042  
Node 135: 0.0042  
Node 205: 0.0042  
Node 191: 0.0041  
Node 237: 0.0041  
Node 257: 0.0041  
Node 262: 0.0040  
Node 295: 0.0040  
Node 45: 0.0039  
Node 123: 0.0039  
Node 12: 0.0037  
Node 107: 0.0037  
Node 73: 0.0035

Node 172: 0.0033  
Node 201: 0.0033  
Node 296: 0.0033  
Node 5: 0.0032  
Node 142: 0.0032  
Node 286: 0.0032  
Node 290: 0.0032  
Node 40: 0.0031  
Node 140: 0.0031  
Node 77: 0.0030  
Node 105: 0.0029  
Node 23: 0.0028  
Node 66: 0.0028  
Node 106: 0.0028  
Node 78: 0.0026  
Node 121: 0.0026  
Node 152: 0.0026  
Node 153: 0.0026  
Node 190: 0.0026  
Node 248: 0.0026  
Node 86: 0.0024  
Node 234: 0.0024  
Node 84: 0.0024  
Node 130: 0.0024  
Node 255: 0.0023  
Node 85: 0.0022  
Node 143: 0.0020  
Node 112: 0.0020  
Node 225: 0.0018  
Node 17: 0.0015  
Node 118: 0.0015  
Node 165: 0.0015  
Node 202: 0.0015  
Node 49: 0.0014  
Node 120: 0.0014  
Node 218: 0.0014  
Node 271: 0.0014  
Node 160: 0.0014  
Node 51: 0.0014  
Node 230: 0.0014  
Node 1: 0.0009  
Node 3: 0.0009  
Node 4: 0.0009  
Node 7: 0.0009  
Node 8: 0.0009

### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity<sup>41</sup>

Node 9: 0.0009  
Node 13: 0.0009  
Node 14: 0.0009  
Node 15: 0.0009  
Node 18: 0.0009  
Node 19: 0.0009  
Node 21: 0.0009  
Node 22: 0.0009  
Node 26: 0.0009  
Node 27: 0.0009  
Node 29: 0.0009  
Node 32: 0.0009  
Node 33: 0.0009  
Node 35: 0.0009  
Node 36: 0.0009  
Node 37: 0.0009  
Node 38: 0.0009  
Node 39: 0.0009  
Node 42: 0.0009  
Node 43: 0.0009  
Node 50: 0.0009  
Node 52: 0.0009  
Node 55: 0.0009  
Node 56: 0.0009  
Node 58: 0.0009  
Node 60: 0.0009  
Node 62: 0.0009  
Node 65: 0.0009  
Node 67: 0.0009  
Node 68: 0.0009  
Node 70: 0.0009  
Node 71: 0.0009  
Node 72: 0.0009  
Node 74: 0.0009  
Node 75: 0.0009  
Node 76: 0.0009  
Node 90: 0.0009  
Node 92: 0.0009  
Node 93: 0.0009  
Node 94: 0.0009  
Node 95: 0.0009  
Node 96: 0.0009  
Node 97: 0.0009  
Node 98: 0.0009  
Node 99: 0.0009

Node 100: 0.0009  
Node 101: 0.0009  
Node 102: 0.0009  
Node 103: 0.0009  
Node 104: 0.0009  
Node 108: 0.0009  
Node 110: 0.0009  
Node 114: 0.0009  
Node 115: 0.0009  
Node 116: 0.0009  
Node 117: 0.0009  
Node 119: 0.0009  
Node 122: 0.0009  
Node 124: 0.0009  
Node 125: 0.0009  
Node 127: 0.0009  
Node 128: 0.0009  
Node 129: 0.0009  
Node 136: 0.0009  
Node 137: 0.0009  
Node 138: 0.0009  
Node 139: 0.0009  
Node 144: 0.0009  
Node 146: 0.0009  
Node 148: 0.0009  
Node 149: 0.0009  
Node 151: 0.0009  
Node 154: 0.0009  
Node 156: 0.0009  
Node 157: 0.0009  
Node 159: 0.0009  
Node 163: 0.0009  
Node 167: 0.0009  
Node 168: 0.0009  
Node 170: 0.0009  
Node 171: 0.0009  
Node 173: 0.0009  
Node 174: 0.0009  
Node 175: 0.0009  
Node 176: 0.0009  
Node 177: 0.0009  
Node 178: 0.0009  
Node 180: 0.0009  
Node 182: 0.0009  
Node 185: 0.0009



### 3.7 Visualisasi Hubungan Antar Kalimat Menggunakan Graf (Graph) berdasarkan Cosine Similarity<sup>43</sup>

Node 188: 0.0009  
Node 189: 0.0009  
Node 192: 0.0009  
Node 193: 0.0009  
Node 200: 0.0009  
Node 203: 0.0009  
Node 204: 0.0009  
Node 209: 0.0009  
Node 212: 0.0009  
Node 214: 0.0009  
Node 215: 0.0009  
Node 217: 0.0009  
Node 220: 0.0009  
Node 221: 0.0009  
Node 222: 0.0009  
Node 223: 0.0009  
Node 224: 0.0009  
Node 228: 0.0009  
Node 233: 0.0009  
Node 236: 0.0009  
Node 240: 0.0009  
Node 242: 0.0009  
Node 245: 0.0009  
Node 247: 0.0009  
Node 249: 0.0009  
Node 250: 0.0009  
Node 252: 0.0009  
Node 253: 0.0009  
Node 259: 0.0009  
Node 261: 0.0009  
Node 263: 0.0009  
Node 265: 0.0009  
Node 266: 0.0009  
Node 267: 0.0009  
Node 268: 0.0009  
Node 270: 0.0009  
Node 272: 0.0009  
Node 273: 0.0009  
Node 274: 0.0009  
Node 276: 0.0009  
Node 279: 0.0009  
Node 280: 0.0009  
Node 282: 0.0009  
Node 283: 0.0009  
Node 284: 0.0009

```
Node 285: 0.0009
Node 287: 0.0009
Node 288: 0.0009
Node 293: 0.0009
Node 294: 0.0009
Node 297: 0.0009
Node 298: 0.0009
Node 299: 0.0009
Node 300: 0.0009
```

```
print("3 Node Tertinggi Page Rank :")
sentence = ""
for node, pagerank in sorted_pagerank[:3]:
    top_sentence = kata[node]
    sentence += top_sentence + ", "
    print(f"Node {node}: Page Rank = {pagerank:.4f}")
    print(f"Kalimat: {top_sentence}")
```

```
3 Node Tertinggi Page Rank :
Node 194: Page Rank = 0.0188
Kalimat: ai
Node 208: Page Rank = 0.0167
Kalimat: proses
Node 251: Page Rank = 0.0163
Kalimat: industri
```

```
news = df['Berita'].iloc[15]
print('Berita yang digunakan : ')
news
```

Berita yang digunakan :

'Begini Peran Teknik Industri dalam Pemanfaatan AI di Era Industri 4.0\n\nKOMPAS.com - Implementasi kecerdasan buatan atau artificial intelligence (AI) dalam industri menjadi salah satu poin

```
print('Kata Kunci :', sentence)
```

Kata Kunci : ai, proses, industri,

# 4

## *Meringkas Berita*

Nama : Muhammad Adam Zaky Jiddyansah

Kelas : Pencarian dan Penambangan Web



## 5

### *Menginstall dan importing modul / libray yang akan digunakan*

```
!pip install seaborn matplotlib
import pandas as pd
import nltk
from nltk.tokenize import sent_tokenize
import re
from nltk.corpus import stopwords
from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer, TfidfTransformer
from sklearn.metrics.pairwise import cosine_similarity
from nltk.tokenize import word_tokenize
from collections import Counter
import networkx as nx
import matplotlib.pyplot as plt
import seaborn as sns
```

```
Requirement already satisfied: seaborn in /usr/local/lib/python3.10/dist-packages (0.12.2)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (3.7.1)
Requirement already satisfied: numpy!=1.24.0,>=1.17 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (1.24.0)
Requirement already satisfied: pandas>=0.25 in /usr/local/lib/python3.10/dist-packages (from seaborn) (1.5.1)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (1.0.7)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (4.22.0)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (1.0.4)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (21.3)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (9.0.1)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib) (3.0.9)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from pandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas) (2022.7)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil) (1.16.0)
```

```
nltk.download("punkt")
nltk.download("stopwords")
```

```
[nltk_data] Downloading package punkt to /root/nltk_data...
```

48            *5 Menginstall dan importing modul / libray yang akan digunakan*

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

## 6

### Membaca Data Berita

```
from google.colab import drive
drive.mount('/content/drive')
file_path = "/content/drive/My Drive/Penambangan WEB/Data/Data_BeritaKompas.csv"
df = pd.read_csv(file_path)
df
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive")

|    | Judul   | Berita                                   |
|----|---|--|
| 0  | Lirik dan Makna Lagu Oh Indang Oh Apang, Lagu ... | Lirik dan Makna Lagu Oh Indang Oh Apa    |
| 1  | Berapa Banyak Kalori yang Dibakar dengan Berja... | Berapa Banyak Kalori yang Dibakar deng   |
| 2  | Pendaftaran Petugas KPPS Pemilu 2024 Dibuka, B... | Pendaftaran Petugas KPPS Pemilu 2024 I   |
| 3  | Produser Ungkap Tiket Advance Siksa Neraka Sol... | Produser Ungkap Tiket Advance Siksa Ne   |
| 4  | Peringati Hari Korban 40 Ribu Jiwa, Pj Gubernu... | Peringati Hari Korban 40 Ribu Jiwa, Pj G |
| 5  | Lirik dan Chord Lagu Salam Kenal - Vidi Aldiano   | Lirik dan Chord Lagu Salam Kenal - Vidi  |
| 6  | Pikap Tabrak Truk Tronton di Kota Malang, Satu... | Pikap Tabrak Truk Tronton di Kota Mala   |
| 7  | Perbandingan Yamaha XMAX 250 Tech Max dan Kymc... | Perbandingan Yamaha XMAX 250 Tech M      |
| 8  | Hasil Mediasi Keluarga Bayi HNM: RS Hermina Po... | Hasil Mediasi Keluarga Bayi HNM: RS He   |
| 9  | Tanggapi Hasil Survei Litbang "Kompas", Mahfud... | Tanggapi Hasil Survei Litbang "Kompas",  |
| 10 | Kepsek SMA 17 Makassar Dimutasi Dampak Aksi De... | Kepsek SMA 17 Makassar Dimutasi Damp     |
| 11 | Status WhatsApp Bakal Bisa Dibagikan Langsung ... | Status WhatsApp Bakal Bisa Dibagikan L   |
| 12 | Alat yang Bekerja Berdasarkan Prinsip Tuas        | Alat yang Bekerja Berdasarkan Prinsip Tu |
| 13 | Pelaku Perusakan Belasan Mobil Dinas di Semara... | Pelaku Perusakan Belasan Mobil Dinas di  |
| 14 | Rafael Alun Dihukum Bayar Uang Pengganti 18,9 ... | Rafael Alun Dihukum Bayar Uang Pengga    |
| 15 | Begini Peran Teknik Industri dalam Pemanfaatan... | Begini Peran Teknik Industri dalam Pema  |
| 16 | Lexus Siapkan RZ550e, RZ Berperforma Tinggi       | Lexus Siapkan RZ550e, RZ Berperforma T   |
| 17 | Pengendara Motor Tewas di Lokasi Kejadian Sete... | Pengendara Motor Tewas di Lokasi Kejad   |
| 18 | Newcastle Vs Milan, Kabar Baik untuk Rossoneri... | Newcastle Vs Milan, Kabar Baik untuk Ro  |

## 6.1 Mengambil salah satu *berita*

```
berita = df['Berita'].iloc[15]  
berita
```

'Begini Peran Teknik Industri dalam Pemanfaatan AI di Era Industri 4.0\n\nKOMPAS.com -  
Implementasi kecerdasan buatan atau artificial intelligence (AI) dalam industri menjadi salah satu poi



# 7

## *Pra-pemrosesan Teks / Preprocessing*

Mengambil teks sebagai input dan melakukan beberapa langkah pra-pemrosesan pada teks dalam bahasa Indonesia. Langkah-langkah tersebut mencakup penghapusan angka, pembersihan karakter khusus, konversi teks ke huruf kecil, penghapusan kata-kata umum (stopwords), dan penggabungan kata-kata yang telah melalui proses sebelumnya. Tujuannya adalah membersihkan dan mempersiapkan teks agar dapat digunakan lebih lanjut dalam analisis atau pemrosesan lanjutan.

```
def preprocessing(text):
    text = re.sub(r'\d+', '', text)
    text = re.sub(r'^\w\s.', '', text)
    text = text.lower()

    stop_words = set(stopwords.words('indonesian'))
    words = text.split()
    filtered_words = [word for word in words if word.lower() not in stop_words]

    preprocessing_text = ' '.join(filtered_words)

    return preprocessing_text

berita = preprocessing(berita)
print(berita)
```

peran teknik industri pemanfaatan ai era industri . kompas.com implementasi kecerdasan buatan artifical

```
kalimat = nltk.sent_tokenize(berita)
kalimat = [sentence.replace('.', ' ') for sentence in kalimat]
print(kalimat)
```

['peran teknik industri pemanfaatan ai era industri ', 'kompascom implementasi kecerdasan buatan artifical



# 8

## Ekstraksi kalimat

Ekstraksi kalimat adalah suatu teknik atau proses yang bertujuan untuk mengidentifikasi dan mengekstrak kalimat-kalimat penting atau relevan dari sebuah dokumen teks. Tujuan utamanya adalah untuk merangkum informasi yang terkandung dalam teks tanpa menghilangkan inti dari pesan yang ingin disampaikan. Metode ekstraksi kalimat berfokus pada pemilihan kalimat-kalimat yang dianggap memiliki nilai informasi tinggi atau representatif dari keseluruhan teks.

### 8.1 Metode pengukuran TF-IDF

Mengonversi teks yang sudah melewati tahap pra-pemrosesan menjadi representasi matriks numerik menggunakan skema TF-IDF. Matriks ini mencerminkan bobot TF-IDF dari setiap kata dalam setiap kalimat. Setelah transformasi, DataFrame pandas dibuat untuk menyajikan matriks tersebut dengan kata-kata sebagai kolom dan setiap baris mewakili satu kalimat, menunjukkan bobot TF-IDF dari kata-kata tersebut dalam kalimat tersebut.

```
tfidf_vectorizer = TfidfVectorizer()
tfidf_preprocessing = tfidf_vectorizer.fit_transform(kalimat)

terms = tfidf_vectorizer.get_feature_names_out()
tfidf_preprocessing = pd.DataFrame(data=tfidf_preprocessing.toarray(), columns=terms)

tfidf_preprocessing
```

|   | ai       | akses    | akurat | alat     | all     | analisis | aplikasi | ar       | artificial | as       |
|---|----------|----------|--------|----------|---------|----------|----------|----------|------------|----------|
| 0 | 0.242502 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.000000   | 0.000000 |
| 1 | 0.142902 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.301414   | 0.000000 |
| 2 | 0.187425 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.000000   | 0.000000 |
| 3 | 0.333926 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.000000   | 0.000000 |
| 4 | 0.000000 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.000000   | 0.000000 |
| 5 | 0.000000 | 0.000000 | 0.0000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.000000   | 0.000000 |

[illegible]

# 9

## Membuat Graph

Membuat graf yang merepresentasikan hubungan antara kalimat-kalimat berdasarkan nilai cosine similarity. Setiap kalimat direpresentasikan sebagai node, dan garis yang menghubungkan dua node menunjukkan tingkat kemiripan antara kalimat tersebut. Semakin panjang dan tebal garis, semakin tinggi tingkat kemiripan antara dua kalimat. Graf ini membantu memvisualisasikan struktur hubungan antar kalimat dalam teks.

### 9.1 Perhitungan Kemiripan Kosinus

Menghitung kemiripan kosinus antara semua pasangan kalimat menggunakan matriks TF-IDF yang telah dihasilkan sebelumnya. Kemudian, hasil kemiripan kosinus tersebut disimpan dalam DataFrame untuk kemudahan analisis.

**Rumus Perhitungan Kemiripan Kosinus:** \* Untuk dua vektor A dan B, kemiripan kosinus dihitung dengan rumus

$$\text{Similarity}(A, B) = \frac{A \cdot B}{\|A\| \cdot \|B\|}$$

- Di sini,  $\cdot$  adalah operasi perkalian dot (dot product), dan  $\|A\|$  adalah norma Euclidean. Dalam konteks ini, vektor A dan B adalah vektor representasi TF-IDF dari dua kalimat. Hasilnya adalah skalar yang menunjukkan seberapa mirip dua kalimat tersebut, dengan nilai 1 menunjukkan kemiripan sempurna.

```
cosine_preprocessing = cosine_similarity(tfidf_preprocessing, tfidf_preprocessing)
```

```
similarity_preprocessing = pd.DataFrame(cosine_preprocessing, columns=range(len(kalimat)), index=range(len(kalimat)))
```

|   | 0        | 1        | 2        | 3        | 4        | 5        | 6        | 7        | 8   | 9        |
|---|----------|----------|----------|----------|----------|----------|----------|----------|-----|----------|
| 0 | 1.000000 | 0.184595 | 0.045451 | 0.431353 | 0.061610 | 0.000000 | 0.149652 | 0.031021 | 0.0 | 0.166220 |
| 1 | 0.184595 | 1.000000 | 0.026783 | 0.215648 | 0.018153 | 0.000000 | 0.060953 | 0.018280 | 0.0 | 0.057353 |

[illegible]

|    | 0        | 1        | 2        | 3        | 4        | 5        | 6        | 7        | 8   | 9        |
|----|----------|----------|----------|----------|----------|----------|----------|----------|-----|----------|
| 45 | 0.000000 | 0.062608 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.0 | 0.000000 |
| 46 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.0 | 0.000000 |

## 9.2 Graph

```
G_preprocessing = nx.DiGraph()

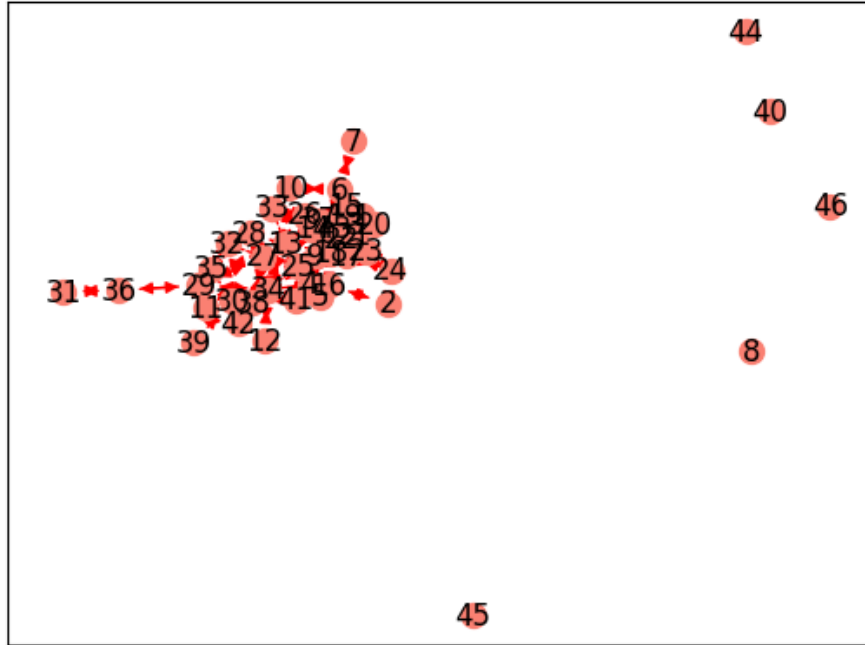
# Add nodes to the graph
for i in range(len(cosine_preprocessing)):
    G_preprocessing.add_node(i)

# Add edges based on similarity
for i in range(len(cosine_preprocessing)):
    for j in range(len(cosine_preprocessing)):
        similarity = cosine_preprocessing[i][j]
        if similarity > 0.1 and i != j:
            G_preprocessing.add_edge(i, j)

# Specify the layout (spring_layout for example)
pos = nx.spring_layout(G_preprocessing)

# Draw the graph
nx.draw_networkx_nodes(G_preprocessing, pos, node_size=100, node_color='salmon')
nx.draw_networkx_edges(G_preprocessing, pos, edge_color='red', arrows=True)
nx.draw_networkx_labels(G_preprocessing, pos)

plt.show()
```





# 10

## *Presentasikan Ukuran dengan Matriks Sentralitas*

### 10.1 Closeness

Menghitung dan menganalisis Closeness Centrality pada graf G\_preprocessing

```
closeness_preprocessing = nx.closeness centrality(G_preprocessing)

sorted_closeness_preprocessing = sorted(closeness_preprocessing.items(), key=lambda x: x[1], reverse=True)
print("Closeness Centrality:")
for node, closeness in sorted_closeness_preprocessing:
    print(f"Node {node}: {closeness:.4f}")
```

Closeness Centrality:

Node 18: 0.4746  
Node 34: 0.4746  
Node 0: 0.4626  
Node 25: 0.4568  
Node 27: 0.4568  
Node 22: 0.4153  
Node 13: 0.4106  
Node 16: 0.4106  
Node 21: 0.4060  
Node 43: 0.3972  
Node 3: 0.3929  
Node 17: 0.3888  
Node 23: 0.3888  
Node 30: 0.3888  
Node 4: 0.3847  
Node 9: 0.3847  
Node 19: 0.3847  
Node 1: 0.3807  
Node 26: 0.3807

Node 41: 0.3767  
Node 29: 0.3729  
Node 37: 0.3691  
Node 15: 0.3654  
Node 14: 0.3583  
Node 5: 0.3548  
Node 32: 0.3548  
Node 35: 0.3447  
Node 20: 0.3384  
Node 6: 0.3353  
Node 38: 0.3322  
Node 33: 0.3292  
Node 42: 0.3263  
Node 11: 0.3234  
Node 24: 0.3206  
Node 10: 0.3178  
Node 12: 0.3123  
Node 28: 0.3097  
Node 2: 0.3020  
Node 39: 0.2727  
Node 36: 0.2687  
Node 7: 0.2453  
Node 31: 0.2076  
Node 8: 0.0000  
Node 40: 0.0000  
Node 44: 0.0000  
Node 45: 0.0000  
Node 46: 0.0000

# 11

## *Hasil Ringkasan Berita Dengan Closeness Centrality*

```
ringkasan_closeness_preprocessing = ""
print("Tiga Node Tertinggi Closeness Centrality Menggunakan Preprocessing:")
for node, closeness_preprocessing in sorted_closeness_preprocessing[:3]:
    top_sentence = kalimat[node]
    ringkasan_closeness_preprocessing += top_sentence + " "
    print(f"Node {node}: Closeness Centrality = {closeness_preprocessing:.4f}")
    print(f"Kalimat: {top_sentence}\n")
```

Tiga Node Tertinggi Closeness Centrality Menggunakan Preprocessing:

Node 18: Closeness Centrality = 0.4746

Kalimat: teknik industri solusi menghadapi kompleksitas transformasi industri didorong ai ilmu teknik

Node 34: Closeness Centrality = 0.4746

Kalimat: mahasiswa memahami menganalisis proses industri manufaktur menyeluruh

Node 0: Closeness Centrality = 0.4626

Kalimat: peran teknik industri pemanfaatan ai era industri

Ringkasan Berita Menggunakan Closeness Centrality:

teknik industri solusi menghadapi kompleksitas transformasi industri didorong ai ilmu teknik industri

```
print("Ringkasan Berita Menggunakan Closeness Centrality:")
print(ringkasan_closeness_preprocessing)
```

Ringkasan Berita Menggunakan Closeness Centrality:

teknik industri solusi menghadapi kompleksitas transformasi industri didorong ai ilmu teknik industri

### 11.1 Menghitung peringkat halaman dengan Pagerank

Rumus Perhitungan Manual PageRank:

- PageRank mengukur pentingnya suatu node dalam sebuah graf berdasarkan seberapa banyak node lain yang mengarah ke node tersebut. Rumusnya adalah sebagai berikut:

$$PR(x) = (1 - d) + d \left( \frac{L(y)}{PR(y)} \right)$$

- $PR(x)$  adalah PageRank dari node x.
- d adalah faktor damping, biasanya diatur sebagai 0,85.
- $PR(y)$  adalah PageRank dari node yang memiliki tautan ke node y.
- $L(y)$  adalah jumlah tautan keluar dari node y.
- Iterasi dilakukan hingga konvergensi atau sejumlah iterasi tertentu. Semakin tinggi nilai PageRank, semakin “penting” node tersebut dalam graf.

```
pagerank_preprocessing = nx.pagerank(G_preprocessing)

sorted_pagerank_preprocessing= sorted(pagerank_preprocessing.items(), key=lambda x: x[1], reverse=True)
print("Page Rank :")
for node, pagerank_preprocessing in sorted_pagerank_preprocessing:
    print(f"Node {node}: {pagerank_preprocessing:.4f}")
```

Page Rank :

```
Node 34: 0.0512
Node 18: 0.0480
Node 0: 0.0442
Node 21: 0.0377
Node 27: 0.0361
Node 30: 0.0336
Node 25: 0.0334
Node 9: 0.0331
Node 22: 0.0320
Node 29: 0.0309
Node 13: 0.0303
Node 3: 0.0287
Node 43: 0.0276
Node 19: 0.0275
Node 26: 0.0259
Node 16: 0.0252
Node 23: 0.0251
Node 1: 0.0250
Node 6: 0.0236
Node 17: 0.0236
Node 37: 0.0219
Node 20: 0.0218
Node 5: 0.0205
Node 14: 0.0204
Node 4: 0.0202
```

Node 35: 0.0195  
Node 32: 0.0193  
Node 38: 0.0191  
Node 15: 0.0185  
Node 41: 0.0176  
Node 24: 0.0174  
Node 36: 0.0170  
Node 42: 0.0153  
Node 33: 0.0138  
Node 11: 0.0115  
Node 10: 0.0112  
Node 2: 0.0108  
Node 31: 0.0107  
Node 28: 0.0107  
Node 39: 0.0076  
Node 7: 0.0075  
Node 12: 0.0071  
Node 8: 0.0035  
Node 40: 0.0035  
Node 44: 0.0035  
Node 45: 0.0035  
Node 46: 0.0035



# 12

## *Hasil Ringkasan dengan Page Rank*

```
ringkasan_pagerank_preprocessing = ""
print("Tiga Node Tertinggi Page Rank Menggunakan Preprocessing:")
for node, pagerank_preprocessing in sorted_pagerank_preprocessing[:3]:
    top_sentence = kalimat[node]
    ringkasan_pagerank_preprocessing += top_sentence + " "
    print(f"Node {node}: Page Rank = {pagerank_preprocessing:.4f}")
    print(f"Kalimat: {top_sentence}\n")
```

Tiga Node Tertinggi Page Rank Menggunakan Preprocessing:

Node 34: Page Rank = 0.0512

Kalimat: mahasiswa memahami menganalisis proses industri manufaktur menyeluruh

Node 18: Page Rank = 0.0480

Kalimat: teknik industri solusi menghadapi kompleksitas transformasi industri didorong ai ilmu teknik

Node 0: Page Rank = 0.0442

Kalimat: peran teknik industri pemanfaatan ai era industri

```
print("Ringkasan Berita Menggunakan Page Rank:")
print(ringkasan_pagerank_preprocessing)
```

Ringkasan Berita Menggunakan Page Rank:

mahasiswa memahami menganalisis proses industri manufaktur menyeluruh teknik industri solusi menghadapi

### 12.1 menggunakan metode eigenvector centrality untuk menghitung sentralitas eigenvector pada graf

Rumus Perhitungan Manual Eigenvector Centrality: \* Eigenvector Centrality mengukur pentingnya suatu node dalam suatu graf berdasarkan pada seberapa banyak node lain yang terhubung ke node tersebut. Rumusnya adalah

sebagai berikut:

$$x_i = \frac{1}{\lambda} \sum_{j=1}^n A_{ij} x_j$$

- $x_i$  adalah eigenvector centrality dari node  $i$
- $A_{ij}$  adalah elemen matriks ketetanggaan yang menunjukkan apakah ada tautan antara node  $i$  dan  $j$
- $\lambda$  adalah nilai eigen (eigenvalue) yang sesuai dengan eigenvector yang dicari.
- Iterasi dilakukan hingga konvergensi atau sejumlah iterasi tertentu. Semakin tinggi nilai eigenvector centrality, semakin “penting” node tersebut dalam graf.

```
eigenvector_preprocessing = nx.eigenvector_centrality(G_preprocessing)

sorted_eigenvector_preprocessing= sorted(eigenvector_preprocessing.items(), key=lambda x: x[1],
print("Eigen Vector :")
for node, eigenvector_preprocessing in sorted_eigenvector_preprocessing:
    print(f"Node {node}: {eigenvector_preprocessing:.4f}")
```

```
Eigen Vector :
Node 0: 0.3588
Node 18: 0.3579
Node 21: 0.2843
Node 22: 0.2628
Node 3: 0.2366
Node 43: 0.2350
Node 9: 0.2343
Node 19: 0.2339
Node 23: 0.2249
Node 25: 0.2062
Node 1: 0.1965
Node 15: 0.1618
Node 34: 0.1588
Node 26: 0.1552
Node 37: 0.1496
Node 20: 0.1494
Node 27: 0.1486
Node 4: 0.1170
Node 6: 0.1109
Node 17: 0.1055
Node 16: 0.1011
Node 5: 0.0917
Node 14: 0.0859
Node 41: 0.0805
```



*12.1 menggunakan metode eigenvector centrality untuk menghitung sentralitas eigenvector pada graf*<sup>67</sup>

Node 24: 0.0790  
Node 13: 0.0679  
Node 30: 0.0670  
Node 29: 0.0553  
Node 38: 0.0408  
Node 33: 0.0383  
Node 35: 0.0380  
Node 32: 0.0357  
Node 28: 0.0335  
Node 42: 0.0331  
Node 11: 0.0266  
Node 2: 0.0223  
Node 10: 0.0222  
Node 12: 0.0197  
Node 7: 0.0137  
Node 39: 0.0083  
Node 36: 0.0070  
Node 31: 0.0009  
Node 8: 0.0000  
Node 40: 0.0000  
Node 44: 0.0000  
Node 45: 0.0000  
Node 46: 0.0000



# 13

## *Hasil Ringkasan menggunakan Eigen Vektor*

```
ringkasan_eigenvector_preprocessing = ""
print("Tiga Node Tertinggi Eigen Vector Menggunakan Preprocessing:")
for node, eigenvector_preprocessing in sorted_eigenvector_preprocessing[:3]:
    top_sentence = kalimat[node]
    ringkasan_eigenvector_preprocessing += top_sentence + " "
    print(f"Node {node}: Page Rank = {eigenvector_preprocessing:.4f}")
    print(f"Kalimat: {top_sentence}\n")
```

Tiga Node Tertinggi Eigen Vector Menggunakan Preprocessing:

Node 0: Page Rank = 0.3588

Kalimat: peran teknik industri pemanfaatan ai era industri

Node 18: Page Rank = 0.3579

Kalimat: teknik industri solusi menghadapi kompleksitas transformasi industri didorong ai ilmu teknik

Node 21: Page Rank = 0.2843

Kalimat: program studi teknik industri binus university menyajikan kurikulum terkini relevan menjembat

```
print("Ringkasan Berita Menggunakan Eigenvector Centrality:")
print(ringkasan_eigenvector_preprocessing)
```

Ringkasan Berita Menggunakan Eigenvector Centrality:

peran teknik industri pemanfaatan ai era industri teknik industri solusi menghadapi kompleksitas trans



# 14

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## *Summary*

---

In summary, this book has no content whatsoever.



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## *References*

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