## 222112058 Feza Raffa Arnanda Penugasan 6

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```
[]: path = "C:/Users/FEZA/My Drive/00. Drive PC/1.STIS/5. Semester 5/Information_
      →Retrieval [IR] P/berita"
    Casefolding
[]: def case_folding(text):
         text = text.lower()
         return text
    Tokenisasi
[]: import nltk
     # nltk.download('punkt') # Download data yang diperlukan untuk tokenisasi
     from nltk.tokenize import word_tokenize
     def tokenisasi(text):
         tokens = word_tokenize(text)
         return tokens
    Eliminasi Stopword
[]: from nltk.corpus import stopwords
     stop_words = set(stopwords.words('Indonesian'))
[]: def eliminasi_stopword(token):
         return [kata for kata in token if kata not in stop_words]
    Stemming
[]: from Sastrawi.Stemmer.StemmerFactory import StemmerFactory
     def stemming_sastrawi(tokens):
         # Membuat stemmer
         factory = StemmerFactory()
         stemmer = factory.create_stemmer()
         return [stemmer.stem(token) for token in tokens]
[]: import re
     import os
     inverted index = {}
     doc_dict = {}
     i = 1
```

```
for filename in os.listdir(path):
    if (filename.endswith('.txt')):
        file_path = os.path.join(path, filename)
        # Ekstrak angka dari nama file menggunakan regular expressions
       match = re.search(r'\d+', filename)
        if match:
            doc_id = match.group() # Mengambil angka dari nama file sebagai_
 →dokumen ID
            with open (file_path, mode='r', encoding='utf-8') as file:
                text = file.read()
                hasil_case_folding = case_folding(text)
                token = tokenisasi(hasil_case_folding)
                token_bersih = eliminasi_stopword(token)
                stemm_token = stemming_sastrawi(token_bersih)
                stemm_token_final = [item for item in stemm_token if item !=__
 🛶''] # membersihkan term kosong pada hasil stemming sebelumnya
                # Menggabungkan hasil stemming menjadi sebuah teks/paragraf
                doc_dict[doc_id] = ' '.join(stemm_token_final)
                for term in set(stemm_token_final): # penggunaan set untuk_
 →mengantisipasi duplikasi term pada sebuah dokumen
                    if term in inverted_index:
                        inverted_index[term].append(doc_id)
                    else:
                        inverted_index[term] = [doc_id]
```

## []: inverted\_index

```
[]: {'signifikan': ['1', '4'],
      'protokol': ['1'],
      'hijau': ['1'],
      '24': ['1'],
      'cek': ['1'],
      'jakarta': ['1', '2', '3', '4', '5'],
      'siti': ['1'],
      'wilayah': ['1', '5'],
      'nadia': ['1'],
      'laku': ['1'],
      'hitung': ['1'],
      'rencana': ['1', '2'],
      'terap': ['1'],
      'masyarakat': ['1'],
      'direktur': ['1'],
      'cegah': ['1'],
      'dr': ['1', '5'],
      'tingkat': ['1', '4'],
      'januari': ['1', '2'],
      'tular': ['1'],
```

```
'langsung': ['1'],
 'zona': ['1'],
 '-': ['1', '2', '3', '4', '5'],
 '3': ['1'],
 'picu': ['1'],
 'longgar': ['1'],
 'kaji': ['1'],
 '34': ['1'],
 'kendali': ['1'],
 'baru': ['1', '3', '4'],
 'giat': ['1'],
 'bebas': ['1'],
 'kemenkes': ['1', '4'],
 'desember': ['1'],
 'perintah': ['1'],
 'sehat': ['1'],
 'batas': ['1'],
 'kab kota': ['1'],
 'mobilitas': ['1'],
 'p2pml': ['1'],
 'menteri': ['1'],
 'sakit': ['1'],
 'ppkm': ['1'],
 'bijak': ['1'],
 'https': ['1', '2', '3', '4', '5'],
 '2': ['1', '3'],
 'tahap': ['1'],
 'health detik com berita-detikhealth d-5816690 wilayah-kamu-sudah-bebas-
covid-19-cek-34-kabkota-zona-hijau-terbaru': ['1'],
 'ri': ['1', '3', '4'],
 'level': ['1'],
 '2021': ['1'],
 'covid-19': ['1', '2', '3', '4', '5'],
 'tarmizi': ['1'],
 'prof': ['2', '3'],
 'lantas': ['2'],
 'djoerban': ['2', '3'],
 'vaksinasi': ['2'],
 'influenza': ['2'],
 'zubairi': ['2', '3'],
 'ada': ['2'],
 'ketua': ['2', '3'],
 'rutin': ['2'],
 'gantung': ['2'],
 'beri': ['2'],
 'tiga': ['2', '3'],
 'satgas': ['2', '3'],
```

```
'dokter': ['2', '3'],
 'idi': ['2', '3'],
 'booster': ['2', '3'],
 'ikat': ['2', '3'],
 'jelas': ['2'],
 'indonesia': ['2', '3', '4'],
 'dosis': ['2', '3'],
 'kait': ['2', '3'],
 '2022': ['2', '3'],
 'health detik com berita-detikhealth d-5816582 vaksin-covid-19-bakal-rutin-
setiap-tahun-tergantung-ini-penjelasannya': ['2'],
 'vaksin': ['2', '3'],
 'pasti': ['2'],
 'turut': ['2'],
 'corona': ['3', '5'],
 '1-2': ['3'],
 'cs': ['3'],
 'bukti': ['3'],
 'strain': ['3'],
 'varian': ['3', '4', '5'],
 'efektivitas': ['3'],
 'singgung': ['3'],
 '1': ['3'],
 'aku': ['3'],
 'turun': ['3', '5'],
 'dasar': ['3'],
 'sebut': ['3'],
 'lawan': ['3'],
 'virus': ['3'],
 'ikut': ['3'],
 'alami': ['3', '4'],
 'jenis': ['3'],
 'pfizer': ['3'],
 'health detik com berita-detikhealth d-5816534 ri-mulai-suntikkan-booster-
di-2022-masihkah-ampuh-lawan-varian-delta-cs': ['3'],
 'suntik': ['3'],
 'riset': ['3'],
 'delta': ['3', '4', '5'],
 'moderna': ['3'],
 'ampuh': ['3'],
 'pakar': ['3'],
 '90': ['4'],
 'alpha': ['4'],
 'alert': ['4'],
 'sulawesi': ['4'],
 'november': ['4'],
 'balitbangkes': ['4'],
```

```
'tambah': ['4'],
 'beta': ['4'],
 'jawa': ['4'],
 'health detik com berita-detikhealth d-5812940 alert-kasus-varian-delta-
covid-19-di-dki-meningkat': ['4'],
 'dki': ['4'],
 'data': ['4'],
 '13': ['4'],
 'utara': ['4'],
 'barat': ['4', '5'],
 '165': ['4'],
 'asal': ['4'],
 '86': ['4'],
 'total': ['4'],
 '1 327': ['4'],
 'medis': ['5'],
 'health detik com berita-detikhealth d-5813949 corona-di-as-mendadak-naik-lagi-
usai-serangan-delta-sempat-mereda': ['5'],
 'catat': ['5'],
 'serang': ['5'],
 'puncak': ['5'],
 'nasihat': ['5'],
 'serikat': ['5'],
 'stabil': ['5'],
 'area': ['5'],
 'timur': ['5'],
 'minggu': ['5'],
 'gedung': ['5'],
 'laut': ['5'],
 '57': ['5'],
 'anthony': ['5'],
 'panas': ['5'],
 'persen': ['5'],
 'reda': ['5'],
 'senin': ['5'],
 'amerika': ['5'],
 'kepala': ['5'],
 'musim': ['5'],
 'fauci': ['5'],
 '15 11 2021': ['5'],
 'dadak': ['5'],
 'pasien': ['5'],
 'as': ['5'],
 'gelombang': ['5'],
 'nasional': ['5'],
 'pasca': ['5'],
 'putih': ['5']}
```

## []: doc\_dict

- []: {'1': 'wilayah bebas covid-19 cek 34 kab kota zona hijau baru jakarta perintah rencana terap laku batas giat masyarakat ppkm level 3 hitung 24 desember 2021 2 januari 2021 menteri sehat ri bijak ppkm level 3 tahap kaji direktur cegah kendali sakit tular langsung p2pml kemenkes ri dr siti nadia tarmizi ppkm level 3 terap covid-19 signifikan picu tingkat mobilitas longgar protokol sehat https health detik com berita-detikhealth d-5816690 wilayah-kamu-sudah-bebas-covid-19-cek-34-kabkota-zona-hijau-terbaru',
  - '2': 'vaksin covid-19 rutin gantung jelas jakarta beri booster dosis tiga vaksin covid-19 indonesia rencana januari 2022 lantas ada vaksinasi covid-19 vaksinasi influenza ketua satgas covid-19 ikat dokter indonesia idi prof zubairi djoerban pasti kait turut vaksin covid-19 booster vaksinasi covid-19 https health detik com berita-detikhealth d-5816582 vaksin-covid-19-bakal-rutin-setiap-tahun-tergantung-ini-penjelasannya',
  - '3': 'ri suntik booster 2022 ampuh lawan varian delta cs jakarta pakar aku vaksin dosis 1-2 alami turun efektivitas varian corona varian delta booster dosis tiga vaksin covid-19 indonesia 2022 jenis vaksin ikut strain virus baru ketua satgas covid-19 ikat dokter indonesia idi prof zubairi djoerban singgung riset kait efektivitas vaksin covid-19 dosis 1 2 sebut dasar riset efektivitas vaksin covid-19 pfizer moderna bukti turun lawan varian delta https health detik com berita-detikhealth d-5816534 ri-mulai-suntikkan-booster-di-2022-masihkah-ampuh-lawan-varian-delta-cs',
  - '4': 'alert varian delta covid-19 dki tingkat jakarta data baru balitbangkes kemenkes ri 13 november tambah varian delta tambah jawa barat 165 dki jakarta 90 sulawesi utara 86 balitbangkes dki jakarta alami tingkat varian delta signifikan varian varian alpha varian delta beta indonesia asal dki jakarta total 1 327 https health detik com berita-detikhealth d-5812940 alert-kasus-varian-delta-covid-19-di-dki-meningkat',
  - '5': 'corona as dadak serang delta reda jakarta covid-19 wilayah amerika serikat as covid-19 catat stabil pasca serang varian delta musim panas kepala nasihat medis gedung putih dr anthony fauci senin 15 11 2021 nasional turun 57 persen minggu puncak gelombang varian delta musim panas pasien covid-19 area barat timur laut dadak https health detik com berita-detikhealth d-5813949 corona-di-as-mendadak-naik-lagi-usai-serangan-delta-sempat-mereda'}

Vocabulary List dari Inverted Index

```
[]: vocab = list(inverted_index.keys())
print(f'Vocabulary List: {vocab}')
```

Vocabulary List: ['signifikan', 'protokol', 'hijau', '24', 'cek', 'jakarta', 'siti', 'wilayah', 'nadia', 'laku', 'hitung', 'rencana', 'terap', 'masyarakat', 'direktur', 'cegah', 'dr', 'tingkat', 'januari', 'tular', 'langsung', 'zona', '-', '3', 'picu', 'longgar', 'kaji', '34', 'kendali', 'baru', 'giat', 'bebas', 'kemenkes', 'desember', 'perintah', 'sehat', 'batas', 'kab kota', 'mobilitas', 'p2pml', 'menteri', 'sakit', 'ppkm', 'bijak', 'https', '2', 'tahap', 'health detik com berita-detikhealth d-5816690 wilayah-kamu-sudah-bebas-

covid-19-cek-34-kabkota-zona-hijau-terbaru', 'ri', 'level', '2021', 'covid-19', 'tarmizi', 'prof', 'lantas', 'djoerban', 'vaksinasi', 'influenza', 'zubairi', 'ada', 'ketua', 'rutin', 'gantung', 'beri', 'tiga', 'satgas', 'dokter', 'idi', 'booster', 'ikat', 'jelas', 'indonesia', 'dosis', 'kait', '2022', 'health detik com berita-detikhealth d-5816582 vaksin-covid-19-bakal-rutin-setiap-tahuntergantung-ini-penjelasannya', 'vaksin', 'pasti', 'turut', 'corona', '1-2', 'cs', 'bukti', 'strain', 'varian', 'efektivitas', 'singgung', '1', 'aku', 'turun', 'dasar', 'sebut', 'lawan', 'virus', 'ikut', 'alami', 'jenis', 'pfizer', 'health detik com berita-detikhealth d-5816534 ri-mulai-suntikkan-boosterdi-2022-masihkah-ampuh-lawan-varian-delta-cs', 'suntik', 'riset', 'delta', 'moderna', 'ampuh', 'pakar', '90', 'alpha', 'alert', 'sulawesi', 'november', 'balitbangkes', 'tambah', 'beta', 'jawa', 'health detik com berita-detikhealth d-5812940 alert-kasus-varian-delta-covid-19-di-dki-meningkat', 'dki', 'data', '13', 'utara', 'barat', '165', 'asal', '86', 'total', '1 327', 'medis', 'health detik com berita-detikhealth d-5813949 corona-di-as-mendadak-naik-lagi-usaiserangan-delta-sempat-mereda', 'catat', 'serang', 'puncak', 'nasihat', 'serikat', 'stabil', 'area', 'timur', 'minggu', 'gedung', 'laut', '57', 'anthony', 'panas', 'persen', 'reda', 'senin', 'amerika', 'kepala', 'musim', 'fauci', '15 11 2021', 'dadak', 'pasien', 'as', 'gelombang', 'nasional', 'pasca', 'putih']

Top 3 Document Retrieval

```
[]: query = 'vaksin corona jakarta'
```

Membuat term frequency

```
'terap': 0,
 'masyarakat': 0,
 'direktur': 0,
 'cegah': 0,
 'dr': 0,
 'tingkat': 0,
 'januari': 0,
 'tular': 0,
 'langsung': 0,
 'zona': 0,
 '-': 0,
 '3': 0,
 'picu': 0,
 'longgar': 0,
 'kaji': 0,
 '34': 0,
 'kendali': 0,
 'baru': 0,
 'giat': 0,
 'bebas': 0,
 'kemenkes': 0,
 'desember': 0,
 'perintah': 0,
 'sehat': 0,
 'batas': 0,
 'kab kota': 0,
 'mobilitas': 0,
 'p2pml': 0,
 'menteri': 0,
 'sakit': 0,
 'ppkm': 0,
 'bijak': 0,
 'https': 0,
 '2': 0,
 'tahap': 0,
 'health detik com berita-detikhealth d-5816690 wilayah-kamu-sudah-bebas-
covid-19-cek-34-kabkota-zona-hijau-terbaru': 0,
 'ri': 0,
 'level': 0,
 '2021': 0,
 'covid-19': 0,
 'tarmizi': 0,
 'prof': 0,
 'lantas': 0,
 'djoerban': 0,
 'vaksinasi': 0,
 'influenza': 0,
```

```
'zubairi': 0,
 'ada': 0,
 'ketua': 0,
 'rutin': 0,
 'gantung': 0,
 'beri': 0,
 'tiga': 0,
 'satgas': 0,
 'dokter': 0,
 'idi': 0,
 'booster': 0,
 'ikat': 0,
 'jelas': 0,
 'indonesia': 0,
 'dosis': 0,
 'kait': 0,
 '2022': 0,
 'health detik com berita-detikhealth d-5816582 vaksin-covid-19-bakal-rutin-
setiap-tahun-tergantung-ini-penjelasannya': 0,
 'vaksin': 1,
 'pasti': 0,
 'turut': 0,
 'corona': 1,
 '1-2': 0,
 'cs': 0,
 'bukti': 0,
 'strain': 0,
 'varian': 0,
 'efektivitas': 0,
 'singgung': 0,
 '1': 0,
 'aku': 0,
 'turun': 0,
 'dasar': 0,
 'sebut': 0,
 'lawan': 0,
 'virus': 0,
 'ikut': 0,
 'alami': 0,
 'jenis': 0,
 'pfizer': 0,
 'health detik com berita-detikhealth d-5816534 ri-mulai-suntikkan-booster-
di-2022-masihkah-ampuh-lawan-varian-delta-cs': 0,
 'suntik': 0,
 'riset': 0,
 'delta': 0,
 'moderna': 0,
```

```
'ampuh': 0,
 'pakar': 0,
 '90': 0,
 'alpha': 0,
 'alert': 0,
 'sulawesi': 0,
 'november': 0,
 'balitbangkes': 0,
 'tambah': 0,
 'beta': 0,
 'jawa': 0,
 'health detik com berita-detikhealth d-5812940 alert-kasus-varian-delta-
covid-19-di-dki-meningkat': 0,
 'dki': 0,
 'data': 0,
 '13': 0,
 'utara': 0,
 'barat': 0,
 '165': 0,
 'asal': 0,
 '86': 0,
 'total': 0,
 '1 327': 0,
 'medis': 0,
 'health detik com berita-detikhealth d-5813949 corona-di-as-mendadak-naik-lagi-
usai-serangan-delta-sempat-mereda': 0,
 'catat': 0,
 'serang': 0,
 'puncak': 0,
 'nasihat': 0,
 'serikat': 0,
 'stabil': 0,
 'area': 0,
 'timur': 0,
 'minggu': 0,
 'gedung': 0,
 'laut': 0,
 '57': 0,
 'anthony': 0,
 'panas': 0,
 'persen': 0,
 'reda': 0,
 'senin': 0,
 'amerika': 0,
 'kepala': 0,
 'musim': 0,
 'fauci': 0,
```

```
'15 11 2021': 0,
'dadak': 0,
'pasien': 0,
'as': 0,
'gelombang': 0,
'nasional': 0,
'pasca': 0,
'putih': 0}
```

Membuat Word Document Frequency

Membuat IDF

```
[]: import numpy as np
def inverseDocFre(vocab, doc_fre, length): # fungsi untuk menghasilkan idf
   idf = {}
   for word in vocab:
      idf[word] = 1 + np.log((length + 1) / (doc_fre[word]+1))
   return idf
```

```
[]: def termFrequencyInDoc(vocab, doc_dict):
    tf_docs = {}
    for doc_id in doc_dict.keys():
        tf_docs[doc_id] = {}
    for word in vocab:
        for doc_id, doc in doc_dict.items():
            tf_docs[doc_id][word] = doc.count(word)
    return tf_docs
```

TF-IDF

```
[]: def tfidf(vocab, tf, idf_scr, doc_dict):
    tf_idf_scr = {}
    for doc_id in doc_dict.keys():
        tf_idf_scr[doc_id] = {}
    for word in vocab:
        for doc_id, doc in doc_dict.items():
            tf_idf_scr[doc_id][word] = tf[doc_id][word] * idf_scr[word]
    return tf_idf_scr
```

```
[]: import math
     def cosine_sim(vec1, vec2):
         vec1 = list(vec1)
         vec2 = list(vec2)
         dot_prod = 0
         for i, v in enumerate(vec1):
             dot prod += v * vec2[i]
         mag_1 = math.sqrt(sum([x**2 for x in vec1]))
         mag_2 = math.sqrt(sum([x**2 for x in vec2]))
         # Menggunakan numpy.asscalar() untuk mengubah array menjadi scalar
         mag 1 = np.squeeze(mag 1)
         mag_2 = np.squeeze(mag_2)
         return dot_prod / (mag_1 * mag_2)
[]: from collections import OrderedDict
     def topk(doc_dict, TD, q, k):
         relevance_scores = {}
         i = 0
         for doc_id in doc_dict.keys():
             relevance_scores[doc_id] = cosine_sim(q, TD[:, i])
             i = i + 1
         sorted_value = OrderedDict(sorted(relevance_scores.items(), key=lambda x:__
      \rightarrow x[1], reverse = True))
         top_k = {j: sorted_value[j] for j in list(sorted_value)[:k]}
         # penghitungan time complexity (disusun oleh banyak dokumen + prosesu
      \rightarrowpengurutan + seleksi top k)
         time_complexity_k = len(doc_dict) + (len(doc_dict) * (len(doc_dict).
      \rightarrowbit_length() - 1)) + k
         return top_k, time_complexity_k
[]: def retriev(vocab, query, doc_dict, k):
         tf_query = termFrequency(vocab, query)
         idf = inverseDocFre(vocab, wordDocFre(vocab, doc_dict), len(doc_dict))
         TQ = np.zeros((len(vocab), 1))
         for word in vocab: # iterasi untuk pemobobotan tf-idf term-query matriks
             ind1 = vocab.index(word) # memberikan index pada tiap kata pada vocab
             TQ[ind1][0] = tf_query[word]*idf[word]
         # implementasi fungsi pembobotan <math>tf-idf antara tiap term dalam vocab dan_{\sqcup}
      →tiap dokumen di dalam corpus untuk digunakan dalam konstruksi term-document
      \rightarrow matriks
         tf_idf = tfidf(vocab, termFrequencyInDoc(vocab, doc_dict), idf, doc_dict)
```

```
# inisialisasi term-query matriks dengan matriks 0 dengan banyak baris⊔
      ⇒sebanyak len(vocab) dan banyak kolom sebanyak len(doc_dict)
         TD = np.zeros((len(vocab), len(doc_dict)))
         for word in vocab: # iterasi untuk konstruksi term-document matriks
             for doc_id, doc in tf_idf.items():
                 ind1 = vocab.index(word)
                 ind2 = list(tf_idf.keys()).index(doc_id)
                 TD[ind1][ind2] = tf_idf[doc_id][word]
         # implementasi fungsi pemilihan top k dokumen beserta penghitungan time_{\sqcup}
      ⇔complexitinya
         top_k_results, complexity_k = topk(doc_dict, TD, TQ, k)
         # penghitungan time complexity (disusun oleh penghitungan tf_query +__
      ⇒pembuatan TQ + penghitungan tf-idf + pembuatan TD + penghitungan top k)
         time complexity main = len(vocab) + len(vocab) + (len(vocab) * |
      -len(doc_dict)) + (len(vocab) * len(doc_dict)) + len(doc_dict) + complexity_k
         return top_k_results, TQ, TD, time_complexity_main
[ ]: k = 3
     top_3_result, TQ, TD, time = retriev(vocab, query, doc_dict, k)
     print(f'Term-query matriks:\n{TQ}')
     print(f'\nTerm-document matriks:\n{TD}')
     print(f'\nHasil perankingan top {k} dokumen:')
     i = 1
     for no_doc, cosine_similarity in top_3_result.items():
         print(f'{i}. Dokumen {no_doc} dengan nilai cosine similarity =__

√{cosine_similarity}')

         i += 1
    Term-query matriks:
    [[0.
                ]
     [0.
                ]
     [0.
                ]
     ГО.
     ГО.
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Hasil perankingan top 3 dokumen:

- 1. Dokumen 2 dengan nilai cosine similarity = [0.22852767]
- 2. Dokumen 3 dengan nilai cosine similarity = [0.1661475]
- 3. Dokumen 5 dengan nilai cosine similarity = [0.07374686]

Time complexity: 1895

## 1 Penugasan Praktikum 6

Evaluasi untuk Unranked Retrieval Set

Mencari skor Precision dan Recall

Dokumen sebenarnya yang sesuai query berdasarkan relevance judgement yaitu berita2 dan berita3.

```
[]: rel_judgement = {
    '1': 0,
    '2': 1,
    '3': 1,
    '4': 0,
    '5': 0,
}

rel_docs = [] # inisilisasi list kosong untuk menyimpan dokumen yang relevan
for doc_id, rel in rel_judgement.items():
    if rel==1:
        rel_docs.append(doc_id)
```

 ${\tt Metrik\ evaluasi\ untuk\ unranked\ retrieval\ set\ adalah\ sebagai\ berikut:}$ 

Nilai precission: 66.6666666666666

Nilai recall: 100.0 Nilai F1-score: 80.0

Pada kode di atas, variabel dictionary rel\_judgement digunakan untuk memberi kode pada berita2 dan berita 3 karena berita tersebut merupakan berita relevan berdasarkan relevance judgment.

Evaluasi untuk Ranked Retrieval Set

```
[]: import numpy as np
     def compute_prf_metrics(I, score, I_Q):
         """Compute precision, recall, F-measures and other
         evaluation metrics for document-level retrieval
         Notebook: C7/C7S3_Evaluation.ipynb
         Args:
             I (np.ndarray): Array of items
             score (np.ndarray): Array containing the score values of the times
             I_Q (np.ndarray): Array of relevant (positive) items
         Returns:
             P_Q (float): Precision
             R_Q (float): Recall
             F_Q (float): F-measures sorted by rank
             BEP (float): Break-even point
             F_max (float): Maximal F-measure
             P_average (float): Mean average
             X_Q (np.ndarray): Relevance function
             rank (np.ndarray): Array of rank values
             I_sorted (np.ndarray): Array of items sorted by rank
             rank_sorted (np.ndarray): Array of rank values sorted by rank
         # Compute rank and sort documents according to rank
         K = len(I)
         index_sorted = np.flip(np.argsort(score))
         I_sorted = I[index_sorted]
         rank = np.argsort(index_sorted) + 1
         rank_sorted = np.arange(1, K+1)
         # Compute relevance function X_Q (indexing starts with zero)
```

```
\# X_Q = np.zeros(K, dtype=bool)
  # for i in range(K):
       if \ I\_sorted[i] \ in \ I\_Q:
            X_Q[i] = True
  X_Q = np.isin(I_sorted, I_Q)
  \# P_Q = np.cumsum(X_Q) / np.arange(1, K+1)
  # Compute precision and recall values (indexing starts with zero)
  M = len(I Q)
  \# P_Q = np.zeros(K)
  \# R_Q = np.zeros(K)
  # for i in range(K):
       r = rank \ sorted[i]
       P_Q[i] = np.sum(X_Q[:r]) / r
       R_Q[i] = np.sum(X_Q[:r]) / M
  P_Q = np.cumsum(X_Q) / np.arange(1, K+1)
  R_Q = np.cumsum(X_Q) / M
  # Break-even point
  BEP = P_Q[M-1]
  # Maximal F-measure
  sum PR = P Q + R Q
  sum_PR[sum_PR == 0] = 1  # Avoid division by zero
  FQ = 2 * (PQ * RQ) / sum PR
  F_{max} = F_{Q.max}()
  # Average precision
  P_average = np.sum(P_Q * X_Q) / len(I_Q)
  return P_Q, R_Q, F_Q, BEP, F_max, P_average, X_Q, rank, I_sorted,_
→rank_sorted
```

```
[]: import pandas as pd
relevance_scores = {}
i = 0
for doc_id in doc_dict.keys():
    relevance_scores[doc_id] = cosine_sim(TQ, TD[:, i])
    i = i + 1

# mengubah value dari dictionary relevance_scores menjadi float
for key, value in relevance_scores.items():
    relevance_scores[key] = float(value[0])

I = np.array(list(relevance_scores.keys()))
score = np.array(list(relevance_scores.values()))
I_Q = np.array(['2', '3'])
output = compute_prf_metrics(I, score, I_Q)
P_Q, R_Q, F_Q, BEP, F_max, P_average, X_Q, rank, I_sorted, rank_sorted = output
```

```
Score $\chi_\mathcal{Q}$
                                           P(r) R(r)
                                                           F(r)
  Rank ID
0
     1 2 0.228528
                                  True 1.000000
                                                  0.5 0.666667
1
     2 3 0.166147
                                  True 1.000000
                                                  1.0 1.000000
2
     3 5 0.073747
                                 False 0.666667
                                                  1.0 0.800000
     4 4 0.051724
                                 False 0.500000
                                                  1.0 0.666667
     5 1 0.011144
                                 False 0.400000
                                                  1.0 0.571429
Break-even point = 1.00
F \max = 1.00
Average precision = 1.0
```

Pada hasil di atas, diperoleh hasil relevance score untuk masing-masing dokumen

```
[]: from matplotlib import pyplot as plt
     def plot_PR_curve(P_Q, R_Q, figsize=(3, 3)):
         fig, ax = plt.subplots(1, 1, figsize=figsize)
         plt.plot(R_Q, P_Q, linestyle='--', marker='o', color='k', mfc='r')
         plt.xlim([0, 1.1])
         plt.ylim([0, 1.1])
         ax.set_aspect('equal', 'box')
         plt.title('PR curve')
         plt.xlabel('Recall')
         plt.ylabel('Precision')
         plt.grid()
         plt.tight_layout()
         ax.plot(BEP, BEP, color='green', marker='o', fillstyle='none',
      ⇒markersize=15)
         ax.set_title('PR curve')
         plt.show()
         return fig, ax
```

```
[]: plot_PR_curve(P_Q, R_Q, figsize=(3,3))
```

