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Lab: 3

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Computing Document Similarity using VSM
         EXERCISE-1
 In [6]: from sklearn.feature_extraction.text import TfidfVectorizer
         import pandas as pd
 In [7]: docs = ["good movie", "not a good movie", "did not like","i like it","good one"]
In [8]: print(docs)
         ['good movie', 'not a good movie', 'did not like', 'i like it', 'good one']
 In [9]: tfidf=TfidfVectorizer(min_df=2,max_df=0.5,ngram_range=(1,2))
In [11]: features=tfidf.fit_transform(docs)
In [12]: print(features)
           (0, 2)
                         0.7071067811865476
           (0, 0)
                         0.7071067811865476
           (1, 2)
                         0.5773502691896257
           (1, 0)
                         0.5773502691896257
           (1, 3)
                         0.5773502691896257
           (2, 3)
                         0.7071067811865476
           (2, 1)
                         0.7071067811865476
           (3, 1)
                         1.0
In [14]: df=pd.DataFrame(features.todense(),columns=tfidf.get_feature_names())
In [15]: print(df)
            good movie
                            like
                                     movie
              0.707107 0.000000 0.707107
                                            0.000000
              0.577350 0.000000
                                  0.577350
                                            0.577350
         1
              0.000000
                       0.707107
                                  0.000000
                                            0.707107
              0.000000 1.000000 0.000000 0.000000
              0.000000
                       0.000000 0.000000 0.000000
         EXERCISE-2
In [23]: tfidf=TfidfVectorizer(min_df=2,max_df=0.8,ngram_range=(2,2))
         features=tfidf.fit_transform(docs)
         print(features)
                         1.0
           (0, 0)
           (1, 0)
                         1.0
         EXERCISE-3
In [24]: | from sklearn.metrics.pairwise import linear_kernel
In [26]: | docl = features[0:1]
         doc2 = features[1:2]
         score = linear_kernel(docl, doc2)
         print(score)
         [[1.]]
In [27]: scores = linear_kernel(docl, features)
         print (scores)
         [[1. 1. 0. 0. 0.]]
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In []:

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In [32]: query = "I like this good movie"
          feature = tfidf.transform([query])
         scores2 = linear_kernel(docl, features)
         print (scores2)
         [[1. 1. 0. 0. 0.]]
         EXERCISE-4
In [30]: | mouse", "the cat saw the mouse", "the mouse ran away from the house", "the cat finally ate the mouse", "the end of the mouse story"]
In [31]: docs
Out[31]: ['the house had a tiny little mouse',
           'the cat saw the mouse',
           'the mouse ran away from the house',
           'the cat finally ate the mouse',
          'the end of the mouse story']
In [33]: tfidf=TfidfVectorizer(min_df=2,max_df=0.5,ngram_range=(1,2))
          features=tfidf.fit_transform(docs)
         print(features)
           (0, 1)
                         0.7071067811865476
           (0, 3)
                         0.7071067811865476
           (1, 0)
                         0.7071067811865476
           (1, 2)
                         0.7071067811865476
           (2, 1)
                         0.7071067811865476
           (2, 3)
                         0.7071067811865476
           (3, 0)
                         0.7071067811865476
                         0.7071067811865476
           (3, 2)
In [34]: | scores_2 = linear_kernel(features[3], features)
In [35]: scores_2
Out[35]: array([[0., 1., 0., 1., 0.]])
In [38]: scores_3 = linear_kernel(features[3], features[0:2])
In [39]: scores_3
Out[39]: array([[0., 1.]])
```