Untitled8

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1 Data Mining and Decision Support

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1.1.1 Singular Value Decomposition

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In [11]: #!/usr/bin/env python
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.datasets import fetch_20newsgroups
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.feature_extraction.text import TfidfTransformer
        from sklearn import decomposition
        corpus = ['To be, or not to be, that is the question',
                  'Whether tis nobler in the mind to suffer',
                  'The slings and arrows of outrageous fortune',
                  'Or to take arms against a sea of troubles',
                  'And by doing something',
                  'the the the the the the'
        ٦
        vectorizer = CountVectorizer(min_df=1)
        dt = vectorizer.fit_transform(corpus)
        x = vectorizer.get_feature_names()
        dt2 = vectorizer.fit_transform(corpus)
        a=dt2.toarray()
        print(a)
[0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 1\ 1\ 0\ 1]
 [0\ 1\ 0\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0]
 [0\ 1\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0]
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In [19]: u ,s , v = np.linalg.svd( dt.toarray () , full_matrices = False )
        b = np.dot (u,np.dot (np.diag (s) , v ))
        #print(b)
        similarity = np.dot(a,b.T)/(np.linalg.norm(a)*np.linalg.norm(b))
        print(similarity)
        # Cosine similarity is simply the cosine of an angle between two given vectors,
        # so it is a number between -1 and 1. If you, however, use it on matrices
        # (as above) and a and b have more than 1 rows,
        # then you will get a matrix of all possible cosines
        # (between each pair of rows between these matrices).
        # So example the result below gives us array of all possible cosines
[[ 1.55555556e-01 3.33333333e-02
                                    1.11111111e-02
                                                    3.3333333e-02
  -1.00228468e-18 7.7777778e-02]
 [ 3.33333333e-02 8.88888889e-02
                                    1.11111111e-02
                                                     1.11111111e-02
  -4.89577514e-18 7.7777778e-02]
 [ 1.11111111e-02 1.11111111e-02
                                    7.7777778e-02
                                                     1.11111111e-02
   1.11111111e-02 7.7777778e-02]
 [ 3.3333333e-02 1.11111111e-02
                                                     8.8888889e-02
                                    1.11111111e-02
   5.74386218e-18 -8.80468538e-17]
 [ -1.33380961e-17 2.55389845e-17
                                    1.11111111e-02
                                                     2.02769900e-17
   4.4444444e-02 -1.06396373e-17]
 [ 7.7777778e-02 7.7777778e-02 7.7777778e-02 -4.39174160e-17
   -4.31753398e-18 5.4444444e-01]]
In []:
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