```
In [1]:
         documents = (
          "The sky is blue",
          "The sun is bright",
          "The sun in the sky is bright",
          "We can see the shining sun, the bright sun"
 In [2]: | documents
 Out[2]: ('The sky is blue',
           'The sun is bright',
           'The sun in the sky is bright',
           'We can see the shining sun, the bright sun')
In [ ]:
         from sklearn.feature_extraction.text import CountVectorizer
In [31]:
In [33]:
         cv =CountVectorizer()
         bow vectorizer= cv.fit transform(documents)
In [35]: | bow_vectorizer.todense()
Out[35]: matrix([[1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0],
                  [0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0],
                  [0, 1, 0, 1, 1, 0, 0, 1, 1, 2, 0],
                  [0, 1, 1, 0, 0, 1, 1, 0, 2, 2, 1]], dtype=int64)
In [45]:
         cv.get_feature_names()
Out[45]: ['blue',
           'bright',
           'can',
           'in',
           'is',
           'see',
           'shining',
           'sky',
           'sun',
           'the',
           'we']
In [53]:
         pd.DataFrame(bow vectorizer.todense(),columns=cv.get feature names(),
                       index=['1st sent', '2nd sent', '3rd sent', '4th sent'])
Out[53]:
                  blue bright can in is see shining sky sun the we
           1st sent
                     1
                           0
                               0
                                  0
                                     1
                                          0
                                                  0
                                                      1
                                                           0
                                                               1
                                                                   0
          2nd sent
                               0
                                  0
                                          0
                                                      0
                     0
                           1
                                     1
                                                           1
                                                               1
                                                                   0
           3rd sent
                                  1
                                                               2
                                                                   0
                           1
                               0
                                          0
                                                           1
```

0

2

2

1

1

1 0 0

1

4th sent

0

1

```
In [41]: | bow_vectorizer[0:1].toarray()
Out[41]: array([[1, 0, 0, 0, 1, 0, 1, 0, 1, 0]], dtype=int64)
In [42]: | print(bow_vectorizer[0:1])
           (0, 9)
                         1
           (0, 7)
                         1
           (0, 4)
                         1
           (0, 0)
                         1
In [ ]:
In [3]:
         from sklearn.feature extraction.text import TfidfVectorizer
In [4]:
         tfidf vectorizer = TfidfVectorizer()
In [29]:
In [5]: | tfidf_matrix = tfidf_vectorizer.fit_transform(documents)
In [11]: | tfidf_matrix
Out[11]: <4x11 sparse matrix of type '<class 'numpy.float64'>'
                 with 21 stored elements in Compressed Sparse Row format>
In [30]: | tfidf_matrix.todense() ##Converted to Dense Matrix
Out[30]: matrix([[0.65919112, 0.
                                        , 0.
                                                   , 0.
                                                                , 0.42075315,
                            , 0.
                                        , 0.51971385, 0.
                  0.
                                                                , 0.34399327,
                  0.
                            ],
                                                 , 0.
                                                                , 0.52210862,
                 [0.
                            , 0.52210862, 0.
                  0.
                            , 0.
                                     , 0.
                                                   , 0.52210862, 0.42685801,
                  0.
                            ],
                 [0.
                            , 0.3218464 , 0. , 0.50423458, 0.3218464 ,
                  0.
                            , 0.
                                       , 0.39754433, 0.3218464 , 0.52626104,
                  0.
                            1,
                           , 0.23910199, 0.37459947, 0.
                  0.37459947, 0.37459947, 0. , 0.47820398, 0.39096309,
                  0.37459947]])
 In [8]: | tfidf_matrix.shape
 Out[8]: (4, 11)
```

```
pd.DataFrame(tfidf_matrix.todense(),columns=tfidf_vectorizer.get_feature_names(),
In [52]:
                         index=['1st sent', '2nd sent', '3rd sent', '4th sent'])
Out[52]:
                            bright
                    blue
                                       can
                                                 in
                                                           is
                                                                        shining
                                                                                              sun
                                                                                                       the
                                                                  see
                                                                                     sky
            1st
                0.659191
                         0.000000 0.000000 0.000000 0.420753 0.000000
                                                                       0.000000
                                                                                         0.000000 0.343993
                                                                                0.519714
                                                                                                           0.00
           sent
           2nd
                0.000000 0.522109
                                  0.000000
                                           0.000000
                                                    0.522109
                                                              0.000000
                                                                       0.000000
                                                                                0.000000
                                                                                         0.522109
                                                                                                  0.426858
                                                                                                            0.00
           sent
            3rd
                0.000000 \quad 0.321846 \quad 0.000000
                                           0.504235
                                                              0.000000
                                                                       0.000000
                                                                                                            0.00
                                                    0.321846
                                                                                0.397544
                                                                                         0.321846
                                                                                                  0.526261
           sent
            4th
                0.000000 0.239102 0.374599
                                           0.000000
                                                    0.000000 0.374599
                                                                      0.374599
                                                                                0.000000
                                                                                        0.478204
                                                                                                  0.390963
In [12]:
          print(tfidf_matrix[0:1])
In [24]:
            (0, 0)
                            0.6591911178676787
            (0, 4)
                            0.42075315164463567
            (0, 7)
                            0.5197138488789809
            (0, 9)
                            0.3439932714296342
In [22]:
          print(tfidf_matrix[0:2])
            (0, 0)
                            0.6591911178676787
             (0, 4)
                            0.42075315164463567
             (0, 7)
                            0.5197138488789809
            (0, 9)
                            0.3439932714296342
            (1, 1)
                            0.5221086219944969
            (1, 8)
                            0.5221086219944969
            (1, 4)
                            0.5221086219944969
            (1, 9)
                            0.42685800978431027
          print(tfidf_matrix[0:3])
In [28]:
            (0, 0)
                            0.6591911178676787
            (0, 4)
                            0.42075315164463567
                            0.5197138488789809
             (0, 7)
            (0, 9)
                            0.3439932714296342
            (1, 1)
                            0.5221086219944969
            (1, 8)
                            0.5221086219944969
            (1, 4)
                            0.5221086219944969
            (1, 9)
                            0.42685800978431027
            (2, 3)
                            0.5042345768555538
            (2, 1)
                            0.32184639875982174
            (2, 8)
                            0.32184639875982174
            (2, 4)
                            0.32184639875982174
            (2, 7)
                            0.3975443320946988
```

(2, 9)

0.5262610401109715

```
In [19]: | print(tfidf_matrix[0:1].toarray())
         [[0.65919112 0.
                                            0.
                                                       0.42075315 0.
                      0.51971385 0.
                                            0.34399327 0.
           0.
                                                                 ]]
         ## Now we have the TF-IDF matrix (tfidf matrix) for each document (the number of
In [43]:
         ## rows of the matrix) with 11 tf-idf terms (the number of columns from the matrix),
         ## we can calculate the Cosine Similarity between the first document("The sky is blue")
         ## with each of the other documents of the set
         from sklearn.metrics.pairwise import cosine similarity
In [20]:
         cosine_similarity(tfidf_matrix[0:1], tfidf_matrix)
In [ ]:
In [ ]:
         array([[ 1.
                            , 0.36651513, 0.52305744, 0.13448867]])
In [ ]: ## The 1st line "The sky is blue" gets a score of 1 with itself | with 2nd line it
         ## gets a score of 0.36 (not similar) with 3rd line it gets a score of 0.52
In [17]:
         import math
         # This was already calculated on the previous step, so we just use the value
         cos sim = 0.52305744
         angle_in_radians = math.acos(cos_sim)
         math.degrees(angle_in_radians)
Out[17]: 58.462437107432784
```

In []: