

# Natural Language Processing

## Data Science Tools 1

Fall 2021



UNIVERSITY *of*  
DENVER

- Common Tasks
- WordNet Corpus
- Word Semantics
- Homework 6 Review



Is it a trivial task to divide a text into sentences?

```
text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We are  
lucky to have him.'"""
```

### **Punkt Sentence Tokenizer**

Unsupervised machine learning model that understands sentence boundaries given a text

Handles Collocations, Abbreviations and many more

```
import nltk.data
text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We are  
lucky to have him.'"""
sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
print('\n-----\n'.join(sent_detector.tokenize(text.strip())))
```

```
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>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
are lucky to have him.'"""
>>> sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
>>> sentences = sent_detector.tokenize(text.strip())
>>> type(sentences)
<class 'list'>
>>> sentences
['Mr. Smith knows how to do his job.', 'He has a Ph.D in Computer Science.', "'We are
lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
```

```
>>> import nltk
>>> tweet = "" "@EmpireStateBldg Central Park Tower is reaaaally hiiiigh""
>>> words = nltk.tokenize.casual.casual_tokenize(tweet,
                                                preserve_case=True,
                                                reduce_len=True,
                                                strip_handles=True)

>>> words
['Central', 'Park', 'Tower', 'is', 'reeaally', 'hiiiigh']
```

```

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>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
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lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], [ "'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
>>> words_pos = [nltk.pos_tag(word_list) for word_list in words]
>>> words_pos
[[('Mr.', 'NNP'), ('Smith', 'NNP'), ('knows', 'VBZ'), ('how', 'WRB'), ('to', 'TO'),
('do', 'VB'), ('his', 'PRP$'), ('job', 'NN'), ('.', '.')], [('He', 'PRP'), ('has',
'VBZ'), ('a', 'DT'), ('Ph.D', 'NNP'), ('in', 'IN'), ('Computer', 'NNP'), ('Science',
'NNP'), ('.', '.')], [("'We", 'NNS'), ('are', 'VBP'), ('lucky', 'JJ'), ('to', 'TO'),
('have', 'VB'), ('him', 'PRP'), ('.', '.'), ("'", "'")]]

```

```

>>> import nltk.data
>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
are lucky to have him.'"""
>>> sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
>>> sentences = sent_detector.tokenize(text.strip())
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>>> sentences
['Mr. Smith knows how to do his job.', 'He has a Ph.D in Computer Science.', "'We are
lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
>>> >>> from nltk.stem.snowball import SnowballStemmer
>>> stemmer = SnowballStemmer('english')
>>> stemmed_words = [[stemmer.stem(word) for word in word_list] for word_list in words]
>>> stemmed_words
[['mr.', 'smith', 'know', 'how', 'to', 'do', 'his', 'job', '.'], ['he', 'has', 'a',
'ph.d', 'in', 'comput', 'scienc', '.'], ['we', 'are', 'lucki', 'to', 'have', 'him', '.',
"'"]]
>>>

```



A similar technique to stemming is **lemmatization**. The difference is that lemmatization provides us with a real word, that is, its canonical form. For example, the lemma of the word *cats* is *cat*, and the lemma for the word *ran* is *run*.

```

>>> import nltk.data
>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
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>>> sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
>>> sentences = sent_detector.tokenize(text.strip())
>>> type(sentences)
<class 'list'>
>>> sentences
['Mr. Smith knows how to do his job.', 'He has a Ph.D in Computer Science.', "'We are
lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
>>> from nltk import WordNetLemmatizer
>>> lemmatizer = WordNetLemmatizer()
>>> words.append(['He', 'ran', 'for', 'mayor', 'in', '1990'])
>>> lemma_words = [[lemmatizer.lemmatize(word, pos='v') for word in word_list] for
word_list in words]
>>> lemma_words
[['Mr.', 'Smith', 'know', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'have', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'be', 'lucky', 'to', 'have', 'him',
'.', "'"], ['He', 'run', 'for', 'mayor', 'in', '1990']]
>>> lemmatizer.lemmatize('worse', 'a')
'bad'
>>> lemmatizer.lemmatize('worse', 'n')
'worse'
>>> lemmatizer.lemmatize('worse', 'v')
'worse'
>>>

```

```
>>> import nltk.data
>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
are lucky to have him.'"""
>>> sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
>>> sentences = sent_detector.tokenize(text.strip())
>>> type(sentences)
<class 'list'>
>>> sentences
['Mr. Smith knows how to do his job.', 'He has a Ph.D in Computer Science.', "'We are
lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
>>> stopwords = nltk.corpus.stopwords.words('english')
>>> type(stopwords)
>>> words_filtered = [word for word_list in words for word in word_list if word not in
stopwords ]
>>> words_filtered
['Mr.', 'Smith', 'knows', 'job', '.', 'He', 'Ph.D', 'Computer', 'Science', '.', "'We",
'lucky', '.', "'", 'He', 'ran', 'mayor', '1990']
```

```

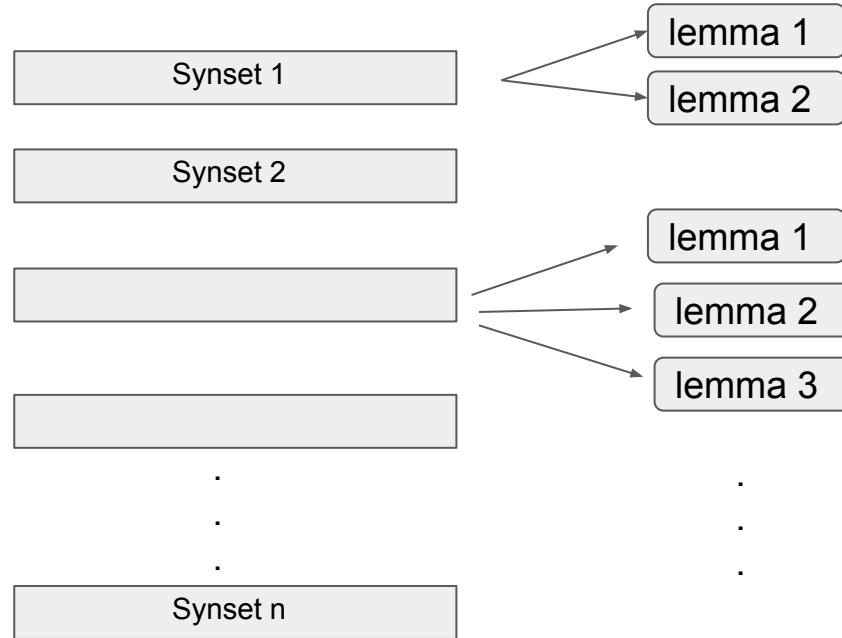
>>> import nltk.data
>>> text = """Mr. Smith knows how to do his job. He has a Ph.D in Computer Science. 'We
are lucky to have him.'"""
>>> sent_detector = nltk.data.load('tokenizers/punkt/english.pickle')
>>> sentences = sent_detector.tokenize(text.strip())
>>> type(sentences)
<class 'list'>
>>> sentences
['Mr. Smith knows how to do his job.', 'He has a Ph.D in Computer Science.', "'We are
lucky to have him.'"]
>>> words = [ nltk.tokenize.word_tokenize(s) for s in sentences]
>>> words
[['Mr.', 'Smith', 'knows', 'how', 'to', 'do', 'his', 'job', '.'], ['He', 'has', 'a',
'Ph.D', 'in', 'Computer', 'Science', '.'], ["'We", 'are', 'lucky', 'to', 'have', 'him',
'.', "'"]]
>>> from nltk.probability import FreqDist
>>> freq = FreqDist(word.lower() for word_list in words for word in word_list)
>>> freq
FreqDist({'.' : 3, 'to' : 2, 'he' : 2, 'in' : 2, 'mr.' : 1, 'smith' : 1, 'knows' : 1, 'how' : 1,
'do' : 1, 'his' : 1, ...})
>>> words_with_frequencies = [(w, f) for w, f in freq.items()]
>>> words_with_frequencies
[('mr.', 1), ('smith', 1), ('knows', 1), ('how', 1), ('to', 2), ('do', 1), ('his', 1),
('job', 1), ('.', 3), ('he', 2), ('has', 1), ('a', 1), ('ph.d', 1), ('in', 2),
('computer', 1), ('science', 1), ("'we", 1), ('are', 1), ('lucky', 1), ('have', 1),
('him', 1), ("'", 1), ('ran', 1), ('for', 1), ('mayor', 1), ('1990', 1)]
>>> stopwords = [t[0] for t in words_with_frequencies if t[1] > 2]
>>> words_filtered_freq = [word for word_list in words for word in word_list if word not
in stopwords ]

```

What is WordNet Corpus?

Digital lexical reference of nouns, adjectives, verbs and adverbs

More than a traditional dictionary or thesaurus



Synonym Sets (synset) - A synset is a unit that has a unique lexical sense. WordNet has a total of 117,000 synsets

## WordNet Search - 3.1

- [WordNet home page](#) - [Glossary](#) - [Help](#)

Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

### Noun

- [S:](#) (n) [dog](#), [domestic dog](#), [Canis familiaris](#) (a member of the genus Canis (probably descended from the common wolf) that has been domesticated by man since prehistoric times; occurs in many breeds) *"the dog barked all night"*
- [S:](#) (n) [frump](#), [dog](#) (a dull unattractive unpleasant girl or woman) *"she got a reputation as a frump"; "she's a real dog"*
- [S:](#) (n) [dog](#) (informal term for a man) *"you lucky dog"*
- [S:](#) (n) [cad](#), [bounder](#), [blackguard](#), [dog](#), [hound](#), [heel](#) (someone who is morally reprehensible) *"you dirty dog"*
- [S:](#) (n) [frank](#), [frankfurter](#), [hotdog](#), [hot dog](#), [dog](#), [wiener](#), [wienerwurst](#), [weenie](#) (a smooth-textured sausage of minced beef or pork usually smoked; often served on a bread roll)
- [S:](#) (n) [pawl](#), [detent](#), [click](#), [dog](#) (a hinged catch that fits into a notch of a ratchet to move a wheel forward or prevent it from moving backward)
- [S:](#) (n) [andiron](#), [firedog](#), [dog](#), [dog-iron](#) (metal supports for logs in a fireplace) *"the andirons were too hot to touch"*

### Verb

- [S:](#) (v) [chase](#), [chase after](#), [trail](#), [tail](#), [tag](#), [give chase](#), [dog](#), [go after](#), [track](#) (go after with the intent to catch) *"The policeman chased the mugger down the alley"; "the dog chased the rabbit"*

## WordNet Search - 3.1

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Word to search for:

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Display options for sense: (gloss) "an example sentence"

**Noun**

- **S: (n) dog, domestic dog, Canis familiaris** (a member of the genus Canis (probably descended from the common wolf) that has been domesticated by man since prehistoric times; occurs in many breeds) *"the dog barked all night"*
- **S: (n) frump, dog** (a dull unattractive unpleasant girl or woman) *"she got a reputation as a frump"; "she's a real dog"*
- **S: (n) dog** (informal term for a man) *"you lucky dog"*
- **S: (n) cad, bounder, blackguard, dog, hound, heel** (someone who is morally reprehensible) *"you dirty dog"*
- **S: (n) frank, frankfurter, hotdog, hot dog, dog, wiener, wienerwurst, weenie** (a smooth-textured sausage of minced beef or pork usually smoked; often served on a bread roll)
- **S: (n) pawl, detent, click, dog** (a hinged catch that fits into a notch of a ratchet to move a wheel forward or prevent it from moving backward)
- **S: (n) andiron, firedog, dog, dog-iron** (metal supports for logs in a fireplace) *"the andirons were too hot to touch"*

**Verb**

- **S: (v) chase, chase after, trail, tail, tag, give chase, dog, go after, track** (go after with the intent to catch) *"The policeman chased the mugger down the alley"; "the dog chased the rabbit"*

```
>>> from nltk.corpus import wordnet as wn
>>> wn.synsets('dog')
[Synset('dog.n.01'), Synset('frump.n.01'),
Synset('dog.n.03'), Synset('cad.n.01'),
Synset('frank.n.02'), Synset('pawl.n.01'),
Synset('andiron.n.01'), Synset('chase.v.01')]
>>> s1 = wn.synsets('dog')[0]
>>> s1
Synset('dog.n.01')
>>> help(s1)
```

```
>>> s1.definition
<bound method Synset.definition of
Synset('dog.n.01')>
>>> s1.definition()
'a member of the genus Canis (probably descended from
the common wolf) that has been domesticated by man
since prehistoric times; occurs in many breeds'
>>> s1.examples()
['the dog barked all night']
>>> s1.lemmas()
[Lemma('dog.n.01.dog'),
Lemma('dog.n.01.domestic_dog'),
Lemma('dog.n.01.Canis_familiaris')]
```

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Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

**Noun**

- [S:](#) (n) [dog](#), [domestic dog](#), [Canis familiaris](#) (a member of the genus Canis (probably descended from the common wolf) that has been domesticated by man since prehistoric times; occurs in many breeds) *"the dog barked all night"*
- [S:](#) (n) [frump](#), [dog](#) (a dull unattractive unpleasant girl or woman) *"she got a reputation as a frump"; "she's a real dog"*
- [S:](#) (n) [dog](#) (informal term for a man) *"you lucky dog"*
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- [S:](#) (n) [frank](#), [frankfurter](#), [hotdog](#), [hot dog](#), [dog](#), [wiener](#), [wienerwurst](#), [weenie](#) (a smooth-textured sausage of minced beef or pork usually smoked; often served on a bread roll)
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**Verb**

- [S:](#) (v) [chase](#), [chase after](#), [trail](#), [tail](#), [tag](#), [give chase](#), [dog](#), [go after](#), [track](#) (go after with the intent to catch) *"The policeman chased the mugger down the alley"; "the dog chased the rabbit"*

```
>>> s4 = wn.synsets('dog')[4]
>>> s4.definition()
'a smooth-textured sausage of minced beef or pork
usually smoked; often served on a bread roll'
>>> s4.lemmas()
[Lemma('frank.n.02.frank'),
Lemma('frank.n.02.frankfurter'),
Lemma('frank.n.02.hotdog'),
Lemma('frank.n.02.hot_dog'), Lemma('frank.n.02.dog'),
Lemma('frank.n.02.wiener'),
Lemma('frank.n.02.wienerwurst'),
Lemma('frank.n.02.weenie')]
```

Lemmas are synonyms of each other in a particular synset. A thesaurus don't have any concept of groupings of words to a synset or lexical sense



## Synsets are linked to other synsets - hypernyms

```

>>> from pprint import pprint
>>> hyp = lambda s:s.hypernyms()
>>> pprint(s1.tree(hyp))
[Synset('dog.n.01'),
 [Synset('canine.n.02'),
  [Synset('carnivore.n.01'),
   [Synset('placental.n.01'),
    [Synset('mammal.n.01'),
     [Synset('vertebrate.n.01'),
      [Synset('chordate.n.01'),
       [Synset('animal.n.01'),
        [Synset('organism.n.01'),
         [Synset('living_thing.n.01'),
          [Synset('whole.n.02'),
           [Synset('object.n.01'),
            [Synset('physical_entity.n.01'),
             [Synset('entity.n.01')]]]]]]]]]]],
 [Synset('domestic_animal.n.01'),
  [Synset('animal.n.01'),
   [Synset('organism.n.01'),
    [Synset('living_thing.n.01'),
     [Synset('whole.n.02'),
      [Synset('object.n.01'),
       [Synset('physical_entity.n.01'), [Synset('entity.n.01')]]]]]]]]]

```

## Synsets are linked to other synsets - hypernyms

```
>>> from pprint import pprint
>>> hyp = lambda s:s.hypernyms()
>>> >>> pprint(s4.tree(hyp))
[Synset('frank.n.02'),
 [Synset('sausage.n.01'),
  [Synset('meat.n.01'),
   [Synset('food.n.02'),
    [Synset('solid.n.01'),
     [Synset('matter.n.03'),
      [Synset('physical_entity.n.01'), [Synset('entity.n.01')]]]]]]]]]
```

## Sunsets are linked to other synsets - hypernyms

```
>>> from pprint import pprint
>>> hyp = lambda s:s.hypernyms()
>>> >>> pprint(s4.tree(hyp))
[Synset('frank.n.02'),
 [Synset('sausage.n.01'),
  [Synset('meat.n.01'),
   [Synset('food.n.02'),
    [Synset('solid.n.01'),
     [Synset('matter.n.03'),
      [Synset('physical_entity.n.01'), [Synset('entity.n.01')]]]]]]]]]

>>> also possible to go the other direction
>>> hyp = lambda s:s.hyponyms()
>>> pprint(s4.tree(hyp))
[Synset('frank.n.02'), [Synset('vienna_sausage.n.01')]]
```

Lemmas are also related - `derivationally_related`, `pertainyms`, `antonyms`

Terms in different syntactic categories that have the same root form and are semantically related.

A relational adjective. Adjectives that are `pertainyms` are usually defined by such phrases as "of or pertaining to" and do not have `antonyms`. A `pertainym` can point to a noun or another `pertainym`.

Lemmas can also have relations between them:

```
>>> vocal = wn.lemma('vocal.a.01.vocal')
>>> vocal.derivationally_related_forms()
[Lemma('vocalize.v.02.vocalize')]
>>> vocal.pertainyms()
[Lemma('voice.n.02.voice')]
>>> vocal.antonyms()
[Lemma('instrumental.a.01.instrumental')]
```

Note: these three relations are only present in Lemmas

## Semantic Similarity between two words

`synset1.path_similarity(synset2)`: Return a score denoting how similar two word senses are, based on the shortest path that connects the senses in the is-a (hypernym/hypnoym) taxonomy.

```
>>> dog = wn.synset('dog.n.01')
>>> cat = wn.synset('cat.n.01')

>>> hit = wn.synset('hit.v.01')
>>> slap = wn.synset('slap.v.01')

>>> dog.path_similarity(cat)
0.2
```

Other similarities functions available from WordNet Interface - Wu-Palmer Similarity, Leacock-Chodorow Similarity, Resnik Similarity, Jiang-Conrath Similarity, Lin Similarity



```
docs = [  
    'the cat sat',  
    'the cat sat in the hat',  
    'the cat with the hat',  
]
```

vocabulary is:

the

cat

sat

in

hat

with

Score words of new sentences based on its  
occurrence in the current document:

hat in the hat

[1, 0, 0, 1, 2, 0]

```
docs = [  
    'the cat sat',  
    'the cat sat in the hat',  
    'the cat with the hat',  
]
```

vocabulary is:

the

cat

sat

in

hat

with

```
>>> docs = [  
...     'the cat sat',  
...     'the cat sat in the hat',  
...     'the cat with the hat',  
... ]  
>>> from sklearn.feature_extraction.text import CountVectorizer  
>>> vectorizer = CountVectorizer()  
>>> X = vectorizer.fit(docs)  
>>> type(X)  
<class 'sklearn.feature_extraction.text.CountVectorizer'>  
>>> new_sentence = "hat in the hat"  
>>> new_sentence_vector = X.transform([new_sentence])  
>>> new_sentence_vector  
<1x6 sparse matrix of type '<class 'numpy.int64'>'  
      with 3 stored elements in Compressed Sparse Row format>  
>>> new_sentence_vector.todense()  
matrix([[0, 2, 1, 0, 1, 0]])
```

Score words of new sentences based on its  
occurrence in the current document:

hat in the hat

[1, 0, 0, 1, 2, 0]



```
docs = [  
    'the cat sat',  
    'the cat sat in the hat',  
    'the cat with the hat',  
]
```

vocabulary is:

the

cat

sat

in

hat

with

Score words of new sentences based on its term frequency times inverse document frequency:

hat in the hat

[1, 0, 0, 1, 2, 0]

Term frequency (tf) = count of t in d / number of words in d

Inverse Document Frequency (idf) =  $\log (N / df) + 1$

TF-IDF = tf \* idf

```
docs = [  
    'the cat sat',  
    'the cat sat in the hat',  
    'the cat with the hat',  
]
```

vocabulary is:

the

cat

sat

in

hat

with

```
>>> docs = [  
...   'the cat sat',  
...   'the cat sat in the hat',  
...   'the cat with the hat',  
... ]  
>>> from sklearn.feature_extraction.text import TfidfVectorizer  
>>> vectorizer = TfidfVectorizer()  
>>> X = vectorizer.fit(docs)  
>>> new_sentence_vector = X.transform([new_sentence])  
>>> new_sentence_vector.todense()  
matrix([[0. , 0.7948031 , 0.52253528, 0. , 0.30861775, 0. ]])
```

Score words of new sentences based on its occurrence in the current document and how rare the word is across documents:

hat in the hat

[1, 0, 0, 1, 2, 0]

```
docs = [
    'the cat sat',
    'the cat sat in the hat',
    'the cat with the hat',
]

vocabulary is:
the
cat
sat
in
hat
with
```

```
>>> from gensim.models import Word2Vec
>>> sentences = [["the", "cat", "sat"], ["the", "cat", "sat", "in", "the", "hat"],
                ["the", "cat", "sat", "with", "the", "hat"]]
>>> model = Word2Vec(sentences, vector_size=6, min_count=1)
>>> model.wv['hat']
array([-0.13808692, -0.15748031,  0.12186277,  0.08450437,  0.11262822,
        0.01271443], dtype=float32)
>>> model.wv['in']
array([-0.12519304, -0.0155007 ,  0.15896864, -0.12198611, -0.03889616,
        -0.0322957 ], dtype=float32)
>>> model.wv['the']
array([-0.00893712,  0.0039405 ,  0.08505583,  0.15015455, -0.15504916,
        -0.11861348], dtype=float32)
```

Score words of new sentences based on its occurrence in the current document and how rare the word is across documents:

hat in the hat

[[...], 0, 0, [...], [...], 0]

```
docs = [
    'the cat sat',
    'the cat sat in the hat',
    'the cat with the hat',
]
```

vocabulary is:

the

cat

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```
>>> from gensim.models import Word2Vec
>>> sentences = [["the", "cat", "sat"], ["the", "cat", "sat", "in", "the", "hat"],
                ["the", "cat", "sat", "with", "the", "hat"]]
>>> model = Word2Vec(sentences, vector_size=6, min_count=1)
>>> model.wv['hat']
array([-0.13808692, -0.15748031,  0.12186277,  0.08450437,  0.11262822,
        0.01271443], dtype=float32)
>>> model.wv['in']
array([-0.12519304, -0.0155007 ,  0.15896864, -0.12198611, -0.03889616,
        -0.0322957 ], dtype=float32)
>>> model.wv['the']
array([-0.00893712,  0.0039405 ,  0.08505583,  0.15015455, -0.15504916,
        -0.11861348], dtype=float32)

>>> import gensim.downloader
>>> print(list(gensim.downloader.info()['models'].keys()))
['fasttext-wiki-news-subwords-300', 'conceptnet-numberbatch-17-06-300',
'word2vec-ruscorpora-300', 'word2vec-google-news-300', 'glove-wiki-gigaword-50',
'glove-wiki-gigaword-100', 'glove-wiki-gigaword-200', 'glove-wiki-gigaword-300',
'glove-twitter-25', 'glove-twitter-50', 'glove-twitter-100', 'glove-twitter-200',
'__testing_word2vec-matrix-synopsis']
```

Score words of new sentences based on its occurrence in the current document and how rare the word is across documents:

hat in the hat

[1, 0, 0, 1, 2, 0]

## Homework 6 Review

