



Word counts with bagof-words

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Bag-of-words

- Basic method for finding topics in a text
- Need to first create tokens using tokenization
- ... and then count up all the tokens
- The more frequent a word, the more important it might be
- Can be a great way to determine the significant words in a text

Bag-of-words example

- Text: "The cat is in the box. The cat likes the box. The box is over the cat."
- Bag of words (stripped punctuation):
 - "The": 3, "box": 3
 - "cat": 3, "the": 3
 - "is": 2
 - "in": 1, "likes": 1, "over": 1



Bag-of-words in Python

```
In [1]: from nltk.tokenize import word_tokenize
In [2]: from collections import Counter
In [3]: Counter(word_tokenize(
                """The cat is in the box. The cat likes the box.
                 The box is over the cat."""))
Out[3]:
Counter({'.': 3,
         'The': 3,
         'box': 3,
         'cat': 3,
         'in': 1,
         'the': 3})
In [4]: counter.most_common(2)
Out[4]: [('The', 3), ('box', 3)]
```





Let's practice!





Simple text preprocessing

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Why preprocess?

- Helps make for better input data
 - When performing machine learning or other statistical methods
- Examples:
 - Tokenization to create a bag of words
 - Lowercasing words
- Lemmatization/Stemming
 - Shorten words to their root stems
- Removing stop words, punctuation, or unwanted tokens
- Good to experiment with different approaches

Preprocessing example

- Input text: Cats, dogs and birds are common pets. So are fish.
- Output tokens: cat, dog, bird, common, pet, fish



Text preprocessing with Python





Let's practice!





Introduction to gensim

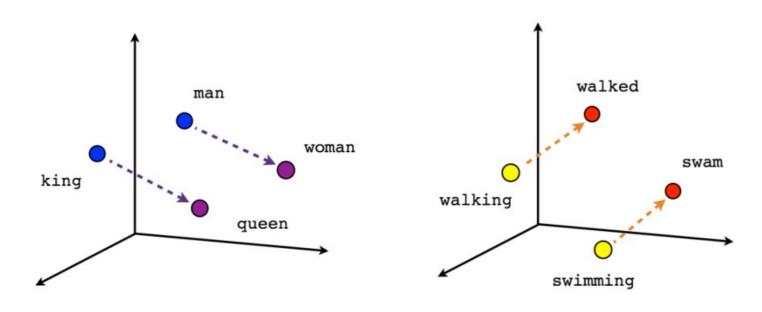
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What is gensim?

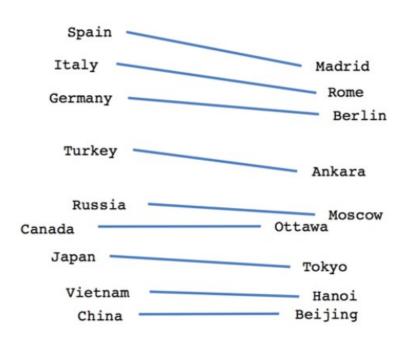
- Popular open-source NLP library
- Uses top academic models to perform complex tasks
 - Building document or word vectors
 - Performing topic identification and document comparison

What is a word vector?



Male-Female

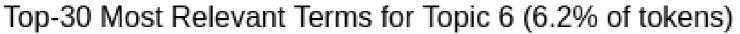
Verb tense

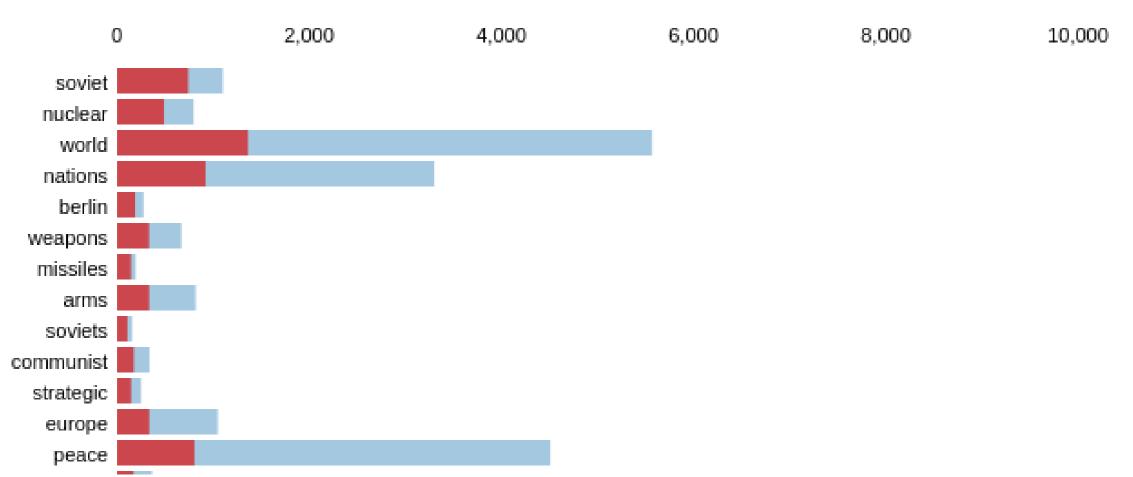


Country-Capital



Gensim Example





(Source: http://tlfvincent.github.io/2015/10/23/presidential-speech-topics)

Creating a gensim dictionary

```
In [1]: from gensim.corpora.dictionary import Dictionary
In [2]: from nltk.tokenize import word_tokenize
In [3]: my_documents = ['The movie was about a spaceship and aliens.',
                        'I really liked the movie!',
   . . . :
                         'Awesome action scenes, but boring characters.',
   . . . :
                         'The movie was awful! I hate alien films.',
   . . . :
                        'Space is cool! I liked the movie.',
   . . . :
                         'More space films, please!',]
   . . . :
In [4]: tokenized docs = [word tokenize(doc.lower())
                           for doc in my documents]
   . . . :
In [5]: dictionary = Dictionary(tokenized docs)
In [6]: dictionary.token2id
Out[6]:
{'!': 11,
 ',': 17,
 '.': 7,
 'a': 2,
 'about': 4,
```

Creating a gensim corpus

```
In [7]: corpus = [dictionary.doc2bow(doc) for doc in tokenized_docs]
In [8]: corpus
Out[8]:
[[(0, 1), (1, 1), (2, 1), (3, 1), (4, 1), (5, 1), (6, 1), (7, 1), (8, 1)],
[(0, 1), (1, 1), (9, 1), (10, 1), (11, 1), (12, 1)],
...
]
```

- gensim models can be easily saved, updated, and reused
- Our dictionary can also be updated
- This more advanced and feature rich bag-of-words can be used in future exercises





Let's practice!





Tf-idf with gensim

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What is tf-idf?

- Term frequency inverse document frequency
- Allows you to determine the most important words in each document
- Each corpus may have shared words beyond just stopwords
- These words should be down-weighted in importance
- Example from astronomy: "Sky"
- Ensures most common words don't show up as key words
- Keeps document specific frequent words weighted high

Tf-idf formula

$$w_{i,j} = tf_{i,j} * \log(rac{N}{df_i})$$

 $w_{i,j} = \text{tf-idf weight for token } i \text{ in document } j$

 $tf_{i,j} = \text{number of occurrences of token } i \text{ in document } j$

 $df_i = \text{number of documents that contain token } i$

N = total number of documents



Tf-idf with gensim

```
In [10]: from gensim.models.tfidfmodel import TfidfModel
In [11]: tfidf = TfidfModel(corpus)

In [12]: tfidf[corpus[1]]
Out[12]:
[(0, 0.1746298276735174),
   (1, 0.1746298276735174),
   (9, 0.29853166221463673),
   (10, 0.7716931521027908),
...
]
```





Let's practice!