Portfolio Assignment: WordNet

1

WordNet is a hierarchal organization of different parts of speech. The purpose of WordNet is to reflect how people organize words in their own minds in the form of a database. Common relationships between words in WordNet include synonymy, hyperonymy, etc)

2

```
In [1]:
         from nltk.corpus import wordnet as wn
         wn.synsets('trunk')
        [Synset('trunk.n.01'),
Out[1]:
         Synset('trunk.n.02'),
         Synset('torso.n.01'),
         Synset('luggage compartment.n.01'),
         Synset('proboscis.n.02')]
        3
In [2]:
         print('Definitions:', wn.synset('luggage compartment.n.01').definition(),'\n')
         print('Usage Examples:', wn.synset('luggage compartment.n.01').examples(),'\n')
         print('Lemmas:', wn.synset('luggage_compartment.n.01').lemmas(),'\n')
        Definitions: compartment in an automobile that carries luggage or shopping or tools
        Usage Examples: ['he put his golf bag in the trunk']
        Lemmas: [Lemma('luggage compartment.n.01.luggage compartment'), Lemma('luggage compartme
        nt.n.01.automobile trunk'), Lemma('luggage compartment.n.01.trunk')]
In [3]:
         trunk = wn.synset('luggage compartment.n.01')
         hyper = lambda s: s.hypernyms()
         list(trunk.closure(hyper))
        [Synset('compartment.n.02'),
Out[3]:
         Synset('room.n.01'),
         Synset('area.n.05'),
         Synset('structure.n.01'),
         Synset('artifact.n.01'),
         Synset('whole.n.02'),
         Synset('object.n.01'),
         Synset('physical entity.n.01'),
         Synset('entity.n.01')]
```

For nouns, the WordNet hierarchy begins with entity, goes down to the type of entity, then continues to go down until it reaches the actual noun itself.

```
In [4]:
         print('Hypernyms:', wn.synset('luggage_compartment.n.01').hypernyms())
         print('Hyponyms:', wn.synset('luggage compartment.n.01').hyponyms())
         print('Meronyms:', wn.synset('luggage_compartment.n.01').part_meronyms())
         print('Holonyms:', wn.synset('luggage_compartment.n.01').part holonyms())
         print('Antonyms:', [1.antonyms() for 1 in wn.synset('luggage compartment.n.01').lemmas()
        Hypernyms: [Synset('compartment.n.02')]
        Hyponyms: [Synset('boot.n.02')]
        Meronyms: []
        Holonyms: [Synset('car.n.01')]
        Antonyms: [[], [], []]
        5
In [5]:
         wn.synsets('lie')
        [Synset('lie.n.01'),
Out[5]:
         Synset('lie.n.02'),
         Synset('lie.n.03'),
         Synset('lie.v.01'),
         Synset('lie.v.02'),
         Synset('dwell.v.02'),
         Synset('lie.v.04'),
         Synset('lie.v.05'),
         Synset('lie.v.06'),
         Synset('lie_down.v.01')]
        6
In [6]:
         print('Definitions:', wn.synset('lie.v.05').definition(),'\n')
         print('Usage Examples:', wn.synset('lie.v.05').examples(),'\n')
         print('Lemmas:', wn.synset('lie.v.05').lemmas(),'\n')
        Definitions: tell an untruth; pretend with intent to deceive
        Usage Examples: ["Don't lie to your parents", 'She lied when she told me she was only 2
        9']
        Lemmas: [Lemma('lie.v.05.lie')]
In [7]:
         trunk = wn.synset('lie.v.05')
         hyper = lambda s: s.hypernyms()
         list(trunk.closure(hyper))
Out[7]: [Synset('misinform.v.01'),
         Synset('inform.v.01'),
         Synset('communicate.v.02'),
         Synset('interact.v.01'),
         Synset('act.v.01')]
        Unlike nouns, there is no top-level synset for verbs. Instead, the hierarchy continues up only until
```

the high possible level that it can.

7

```
In [8]: wn.morphy('lie', wn.VERB)
Out[8]: 'lie'
```

8

```
from nltk.wsd import lesk

cougar = wn.synset('cougar.n.01')
panther = wn.synset('panther.n.02')

print('Wu-Palmer Similarity:', wn.wup_similarity(cougar, panther))

sent = ['Cougars are smaller than lions, but are still dangerous']
print('Lesk:', lesk(sent, 'panther'))
```

Wu-Palmer Similarity: 0.8125 Lesk: Synset('panther.n.02')

The Wu-Palmer Similarity metric and the Lesk algorithm are both effective at capturing the similarity of words.

9

SentiWord is a resource built for sentiment analysis. The way it works is that it assigns positive, negative, and objective scores to synsets from WordNet that can be used to understand someone's opinion.

```
import nltk
nltk.download('sentiwordnet')
from nltk.corpus import sentiwordnet as swn

hate_list = swn.senti_synsets('hate')
for hate in hate_list:
    print(hate)
    print('Positive Score:', hate.pos_score())
    print('Negative Score:', hate.neg_score())
    print('Objective Score:', hate.obj_score())
    print('\n')
```

```
<hate.v.01: PosScore=0.0 NegScore=0.75>
Positive Score: 0.0
Negative Score: 0.75
Objective Score: 0.25
```

```
In [11]:
          sent = 'I hate Indiana Jones'
          tokens = sent.split()
          for token in tokens:
              syn_list = list(swn.senti_synsets(token))
              if syn_list:
                  print(syn_list[0])
                  print('Positive Score:', syn_list[0].pos_score())
                  print('Negative Score:', syn_list[0].neg_score())
                  print('Objective Score:', syn_list[0].obj_score())
                  print('\n')
         <iodine.n.01: PosScore=0.0 NegScore=0.0>
         Positive Score: 0.0
         Negative Score: 0.0
         Objective Score: 1.0
         <hate.n.01: PosScore=0.125 NegScore=0.375>
         Positive Score: 0.125
         Negative Score: 0.375
         Objective Score: 0.5
         <indiana.n.01: PosScore=0.0 NegScore=0.0>
         Positive Score: 0.0
         Negative Score: 0.0
         Objective Score: 1.0
         <jones.n.01: PosScore=0.0 NegScore=0.0>
         Positive Score: 0.0
         Negative Score: 0.0
         Objective Score: 1.0
```

WordNet assigns scores only to synsets that are commonly used in opinionated ways; otherwise, default scores are used. One application for these scores is summing up all the scores in a sentence in order to understand the overall polarity of the sentence.

10

A collocation is a set of words that are commonly found together, despite sometimes having very different meanings. For example "right" and "now" mean very different things, but "right now" is commonly said.

```
In [12]: from nltk.book import *
    text4.collocations()
```

```
*** Introductory Examples for the NLTK Book ***
Loading text1, ..., text9 and sent1, ..., sent9
Type the name of the text or sentence to view it.
Type: 'texts()' or 'sents()' to list the materials.
text1: Moby Dick by Herman Melville 1851
text2: Sense and Sensibility by Jane Austen 1811
text3: The Book of Genesis
text4: Inaugural Address Corpus
text5: Chat Corpus
text6: Monty Python and the Holy Grail
text7: Wall Street Journal
text8: Personals Corpus
text9: The Man Who Was Thursday by G . K . Chesterton 1908
United States; fellow citizens; years ago; four years; Federal
Government; General Government; American people; Vice President; God
bless; Chief Justice; one another; fellow Americans; Old World;
Almighty God; Fellow citizens; Chief Magistrate; every citizen; Indian
tribes; public debt; foreign nations
```

```
In [13]:
```

```
import math
vocab = len(set(text4))
hg = text.count('United States')/vocab
print("p(United States) = ",hg )
h = text.count('United')/vocab
print("p(United) = ", h)
g = text.count('States')/vocab
print('p(States) = ', g)
pmi = math.log2(hg / (h * g))
print('pmi = ', pmi)
```

```
p(United States) = 0.015860349127182045
p(United) = 0.0170573566084788
p(States) = 0.03301745635910224
pmi = 4.815657649820885
```

While simple, the formula for point-wise mutual information is very effective and can tell us which words are likely to be collocations.