

# **Natural Language Processing**

03: Senses and WordNet

Philipp Schaer, Technische Hochschule Köln, Cologne, Germany

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Technology Arts Sciences TH Köln

# Polysemy

- ... the movie is so bad, in fact, that it retains that ridiculous tarzan call that was so tirelessly mocked in last summer's comedy "george of the jungle."
- ... the performances are bad.



# **Polysemy**

	Sense 1	Sense 2
bad	Of poor quality or little worth.	slang (orig. U.S.). Formidable, good.
sick	Suffering from illness of any kind; ill, unwell, ailing.	slang (now esp. Skateboarding sick and Surfing). Excellent, impressive; risky.
awesome	Inspiring awe; appalling, dreadful, weird.	Expressing enthusiastic approval: great, excellent, highly impressive; fantastic.
wicked	Bad in moral character, disposition, or conduct	Excellent, splendid; remarkable. slang (orig. U.S.).

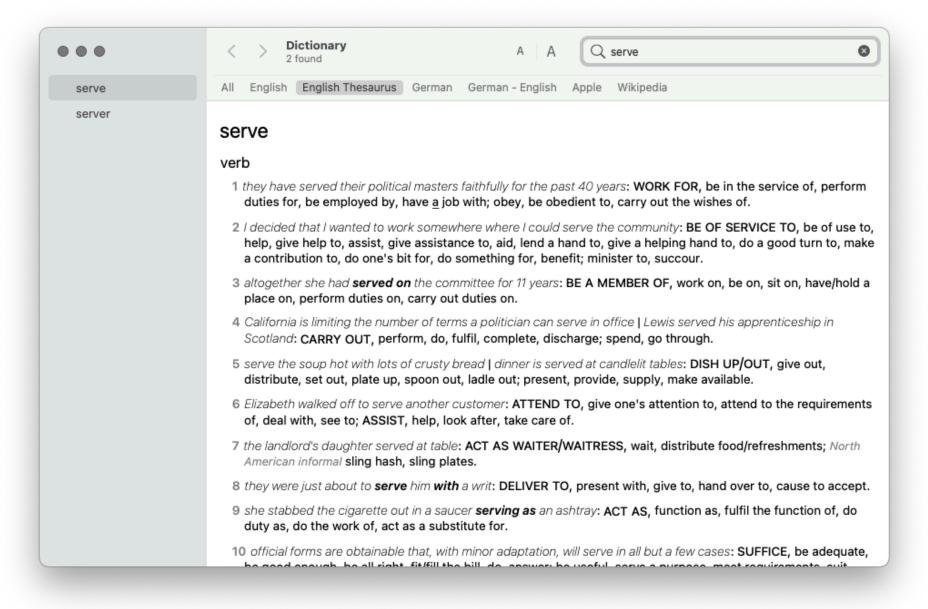
### Word senses

I'm going to the bank

- bank<sub>1</sub> = "financial institution"
- bank<sub>2</sub> = "sloping mound"
- bank<sub>3</sub> = "biological repository"
- bank<sub>4</sub> = "building where a bank<sub>1</sub> does its business"

A word sense is a representation of one **aspect** of a word's meaning.

## Word senses - "serve"



## **Word senses**

- They rarely serve red meat
- He served as U.S. ambassador to Norway
- He might have served his time.

# Zeugma

- Conjunction ("yoke") of antagonistic readings
- one test for whether word senses are distinct.
  - Which flights serve breakfast?
  - Does Midwest Express serve Philadelphia?
  - Does Midwest Express serve breakfast and Philadelphia?

# Relationship between senses

- Synonymy
- Antonymy
- Hypernymy
- Meronymy
- And many others... like Metonymy

# **Synonym**

- Two senses of different words are synonyms of each other if their meaning is nearly identical
- Two words are never exactly the same in their meaning, distribution of use, dialect or other contexts in which they're licensed.
- Synonyms can be exchanged for each other without changing the truth conditions of a sentence.
- Synonymy holds between word senses, not words

couch	sofa
filbert	hazelnut
car	automobile
fair	impartial
fair	pale

# **Antonymy**

- Two senses of different words are antonyms of each other if their meaning is nearly opposite
- All aspects of meaning are nearly identical between antonyms, except one (very much like synonyms in this respect)

long	short	both describe length
big	little	both describe size
fast	slow	both describe speed
cold	hot	both describe temperature
dark	light	both describe luminescence

# **Hyponymy**

hypo = "under" (e.g., hypothermia)

- Sense A is a hyponym of sense B if A is a subclass of B
- Formally, entailment: for entity x,  $A(x) \Rightarrow B(x)$
- Hyponymy is generally transitive
  - dog -> mammal -> animal

hyponym/subordinate	hypernym/superordinate
car	vehicle
mango	fruit
chair	furniture
dog	mammal
mammal	animal

# Meronymy

Part-whole relations. A meronym is a part of a holonym.

meronym	holonym
leg	chair
wheel	car
car	automobile

## WordNet

- Lexical database for nouns, verbs and adjectives/abverbs.
- Each word sense is arranged in a synset (category of nearsynonyms) and each synset is related to others in terms of their sense relations.
- General information: https://wordnet.princeton.edu
- Online interface: <a href="http://wordnetweb.princeton.edu/perl/webwn">http://wordnetweb.princeton.edu/perl/webwn</a>

## **Relations**

Also Called	Definition	Example
Superordinate	From concepts to superordinates	$breakfast^1 \rightarrow meal^1$
Subordinate	From concepts to subtypes	$meal^1  ightarrow lunch^1$
nym Instance	From instances to their concepts	$Austen^1 \rightarrow author^1$
nym Has-Instance	From concepts to their instances	$composer^1 \rightarrow Bach^1$
Has-Part	From wholes to parts	$table^2  ightarrow leg^3$
Part-Of	From parts to wholes	$course^7 \rightarrow meal^1$
	Semantic opposition between lemmas	$leader^1 \iff follower^1$
	Lemmas w/same morphological root	$destruction^1 \iff destroy$
	Superordinate Subordinate rnym Instance nym Has-Instance Has-Part	Superordinate From concepts to superordinates Subordinate From concepts to subtypes From instances to their concepts From concepts to their instances From wholes to parts From parts to wholes Semantic opposition between lemmas

**Figure 18.3** Some of the noun relations in WordNet.

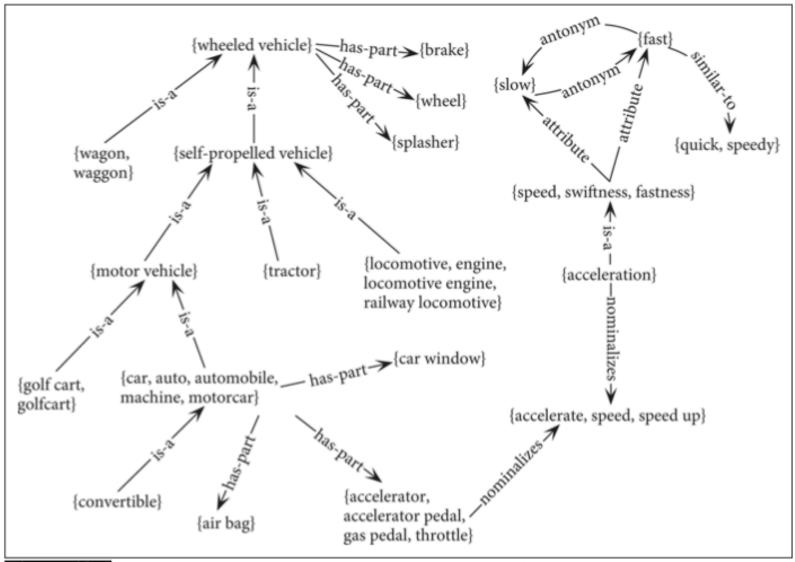
# **Synsets**

synset	gloss
mark, grade, score	a number or letter indicating quality
scratch, scrape, scar, mark	an indication of damage
bell ringer, bull's eye, mark, home run	something that exactly succeeds in achieving its goal
hump, fool, gull, mark, patsy, fall, guy, sucker, soft touch, mug	a person who is gullible and easy to take advantage of
mark, stigma, brand, stain	a symbol of disgrace or infamy

# Hypernyms of bass

```
bass<sup>3</sup>. basso (an adult male singer with the lowest voice)
=> singer, vocalist, vocalizer, vocaliser
   => musician, instrumentalist, player
      => performer, performing artist
         => entertainer
            => person, individual, someone...
               => organism, being
                   => living thing, animate thing,
                      => whole, unit
                         => object, physical object
                            => physical entity
                               => entity
bass<sup>7</sup> (member with the lowest range of a family of instruments)
=> musical instrument, instrument
   => device
      => instrumentality, instrumentation
         => artifact, artefact
            => whole, unit
               => object, physical object
                   => physical entity
                      => entity
```

Figure 18.5 Hyponymy chains for two separate senses of the lemma *bass*. Note that the chains are completely distinct, only converging at the very abstract level *whole*, *unit*.



**Figure 18.6** WordNet viewed as a graph. Figure from Navigli (2016).

## **WordNet**

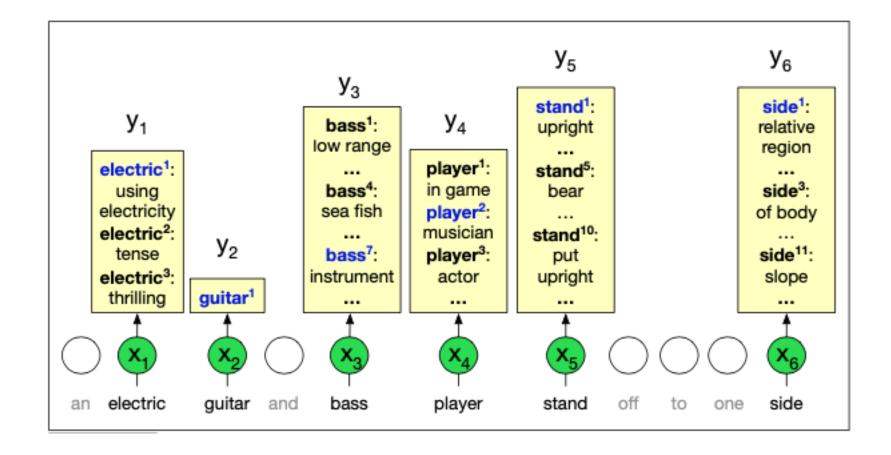
- WordNet encodes human-judged measures of similarity.
- The WordNet 3.0 release has 117,798 nouns, 11,529 verbs, 22,479 adjectives, and 4,481 adverbs. The average noun has 1.23 senses, and the average verb has 2.16 senses.
- By indexing word senses, we can build annotated resources on top of it for word sense disambiguation.

# Word sense disambiguation

#### "All-word" WSD

- "Only<sub>only1</sub> a relative<sub>relative1</sub> handful<sub>handful1</sub> of such<sub>such0</sub> reports<sub>report3</sub> was received<sub>receive2</sub>"
- For all content words in a sentence, resolve each token to its sense in an fixed sense inventory (e.g., WordNet).
- Simple but strong heuristics:
  - Choose the most frequent sense (equivalent, in WordNet, to take the first sense)
  - One sense per discourse

# All-word WSD example



# **Dictionary WSD**

 Predict the sense a given token that has the sense's dictionary highest overlap between the token's context and sense's dictionary gloss.

bank	Gloss Examples	a financial institution that accepts deposits and channels the money into lending activities "he cashed a check at the bank", "that bank holds the mortage on my home"
bank	Gloss Examples	sloping land (especially the slope beside a body of water) "the pulled the canie up on the bank", "he sat on the bank of the river and watched the currents"

"The boat washed up on the river bank."

## **Lesk Algorithm**

```
function SIMPLIFIED LESK(word, sentence) returns best sense of
word
  best-sense ← most frequent sense for word
  max-overlap \leftarrow 0
  context \leftarrow set of words in sentence
  for each sense in senses of word do
    signature ← set of words in gloss and examples of sense
    overlap←COMPUTEOVERLAP(signature,context)
    if overlap > max-overlap then
      max-overlap \leftarrow overlap
      best-sense←sense
  end
  return(best-sense)
```

# **Lesk Algorithm**

- Knowledge-based algorithms, rely solely on WordNet or other such resources and don't require labeled data
- Simplified Lesk algorithm.
  - The COMPUTEOVERLAP function returns the number of words in common between two sets, ignoring function words or other words on a stop list.
  - The original Lesk algorithm defines the context in a more complex way.
- There are many obvious extensions to Simplified Lesk, such as weighing the overlapping words by IDF to downweight frequent words like function words

# Other Applications of WordNet

- ✓ word-sense disambiguation
- information retrieval
- automatic text classification
- automatic text summarization
- machine translation
- and even automatic crossword puzzle generation

A common use of WordNet is to check similarity between words

- measure distance between synsets in WordNet's graph
- NTLK: synset1.path\_similarity(synset2)
- https://www.nltk.org/howto/wordnet.html

### Mental homework

- Using your favourite dictionary, simulate the original Lesk word overlap disambiguation algorithm on the phrase *Time flies like* an arrow. Assume that the words are to be disambiguated one at a time, from left to right, and that the results from earlier decisions are used later in the process.
- How can we use WordNet in the ESUPOL scenario?