

Lexical Semantics

COMP-599

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Outline

Semantics

Lexical semantics

Lexical semantic relations

WordNet

Word Sense Disambiguation

- Lesk algorithm
- Yarowsky's algorithm

Semantics

The study of **meaning** in language

What does meaning mean?

- Relationship of linguistic expression to the real world
- Relationship of linguistic expressions to each other

Let's start by focusing on the meaning of **words**—**lexical semantics**.

Later on:

- meaning of phrases and sentences
- how to construct that from meanings of words

From Language to the World

What does *telephone* mean?

- Picks out all of the objects in the world that are telephones (its **referents**)

Its **extensional** definition



not telephones



Relationship of Linguistic Expressions

How would you define *telephone*? e.g, to a three-year-old, or to a friendly Martian.

Dictionary Definition

<http://dictionary.reference.com/browse/telephone>

Its **intensional definition**

- The necessary and sufficient conditions to be a telephone

This presupposes you know what “apparatus”, “sound”, “speech”, etc. mean.

Sense and Reference (Frege, 1892)

Frege was one of the first to distinguish between the **sense** of a term, and its **reference**.

Same referent, different senses:

Venus →

the morning star →

the evening star →



Lexical Semantic Relations

How specifically do terms relate to each other? Here are some ways:

Hypernymy/hyponymy

Synonymy

Antonymy

Homonymy

Polysemy

Metonymy

Synecdoche

Holonymy/meronymy

Hypernymy/Hyponymy

ISA relationship

Hyponym

monkey

Montreal

red wine

Hypernym

mammal

city

beverage

Synonymy and Antonymy

Synonymy

(Roughly) same meaning

offspring descendent spawn

happy joyful merry

Antonymy

(Roughly) opposite meaning

synonym antonym

happy sad

descendant ancestor

Homonymy

Same form, different (and unrelated) meaning

Homophone – same sound

- e.g., *son* vs. *sun*

Homograph – same written form

- e.g., *lead* (noun) vs. *lead* (verb)

Polysemy

Multiple related meanings

S: (n) **newspaper**, paper (a daily or weekly publication on folded sheets; contains news and articles and advertisements) *"he read his newspaper at breakfast"*

S: (n) **newspaper**, paper, newspaper publisher (a business firm that publishes newspapers) *"Murdoch owns many newspapers"*

S: (n) **newspaper**, paper (the physical object that is the product of a newspaper publisher) *"when it began to rain he covered his head with a newspaper"*

S: (n) **newspaper**, newsprint (cheap paper made from wood pulp and used for printing newspapers) *"they used bales of newspaper every day"*

Homonymy vs Polysemy

Homonymy: unrelated Polysemy: related meaning

S: (n) **position**, place (the particular portion of space occupied by something) *"he put the lamp back in its place"*

S: (n) military position, **position** (a point occupied by troops for tactical reasons)

S: (n) **position**, view, perspective (a way of regarding situations or topics etc.) *"consider what follows from the positivist view"*

S: (n) **position**, posture, attitude (the arrangement of the body and its limbs) *"he assumed an attitude of surrender"*

S: (n) status, **position** (the relative position or standing of things or especially persons in a society) *"he had the status of a minor"; "the novel attained the status of a classic"; "atheists do not enjoy a favorable position in American life"*

S: (n) **position**, post, berth, office, spot, billet, place, situation (a job in an organization) *"he occupied a post in the treasury"*

Metonymy

Substitution of one entity for another related one

We ordered many delicious dishes at the restaurant.

I worked for the local paper for five years.

Quebec City is cutting our budget again.

The loonie is at a 11-year low.

Synecdoche – a specific kind of metonymy involving whole-part relations

All hands on deck!

Don't be a <censored body part>

Holonymy/meronymy

Some kind of whole/part relationship

Subtypes

Holonym

Meronym

groups and members

class

student

whole and part

car

windshield

whole and substance

chair

wood

Quiz

Classify the following examples in terms of what lexical semantic relation they exhibit

cold *freezing*

they're *their*

hair *head*

enemy *friend*

cut (hair) *cut (bread)*

George Clooney *actor*

WordNet (Miller et al., 1990)

WordNet is a lexical resource organized by **synsets**

- Nodes: synsets
- Edges: lexical semantic relation between two synsets

Separate hierarchy for different parts of speech

- Nouns, verbs, adjectives, adverbs

A Synset Entry

S: (n) **hand**, manus, mitt, paw (the (prehensile) extremity of the superior limb) *"he had the hands of a surgeon"; "he extended his mitt"*

direct hyponym / full hyponym

S: (n) fist, clenched fist (a hand with the fingers clenched in the palm (as for hitting))

S: (n) hooks, meat hooks, maulers (large strong hand (as of a fighter)) *"wait till I get my hooks on him"*

S: (n) right, right hand (the hand that is on the right side of the body) *"he writes with his right hand but pitches with his left"; "hit him with quick rights to the body"*

S: (n) left, left hand (the hand that is on the left side of the body) *"jab with your left"*

part meronym

direct hypernym / inherited hypernym / sister term

part holonym

S: (n) arm (a human limb; technically the part of the superior limb between the shoulder and the elbow but commonly used to refer to the whole superior limb)

S: (n) homo, man, human being, human (any living or extinct member of the family Hominidae characterized by superior intelligence, articulate speech, and erect carriage)

derivationally related form

<http://wordnetweb.princeton.edu/perl/webwn?o2=&o0=1&o8=1&o1=1&o7=&o5=&o9=&o6=&o3=&o4=&s=hand&i=8&h=110000000000000000000000000000#>

C

WordNet Has an NLTK Interface

```
>>> from nltk.corpus import wordnet
```

Some useful functions:

```
>>> wordnet.synsets(<query_term>)
```

```
>>> wordnet.synset(<synset_name>)
```

Remember you can use `dir` and `help` to get a list of functions in Python.

Word Sense Disambiguation

Figuring out which word sense is expressed in context

*His **hands** were tired from hours of typing.*

→ *hand.n.01*

*Due to her superior education, her **hand** was flowing and graceful.*

→ *hand.n.03*

General idea: use words in the context to disambiguate.
Which words above would help with this?

Possible Computational Approaches

A heuristic algorithm

- **Lesk's algorithm**

Supervised machine learning

- Possible, but requires a lot of work to annotate word sense information that we want to avoid

Unsupervised, or minimally supervised machine learning

- **Yarowsky's algorithm**

Lesk's Algorithm (1986)

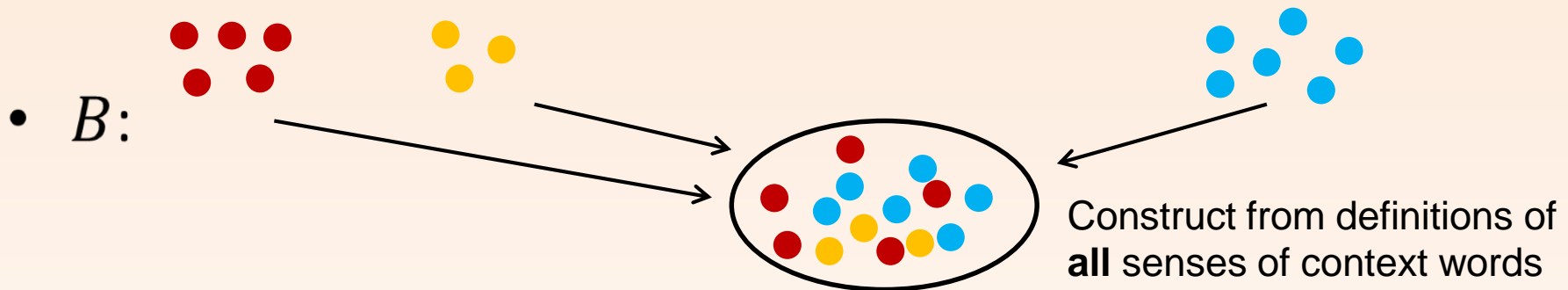
Use the dictionary definitions of a word's senses

Steps to disambiguate word w :

1. Construct a bag of words representation of the context, B
2. For each candidate sense s_i of word w :
 - Calculate a signature of the sense by taking all of the words in the dictionary definition of s_i
 - Compute $\text{Overlap}(B, \text{signature}(s_i))$
3. Select the sense with the highest overlap score

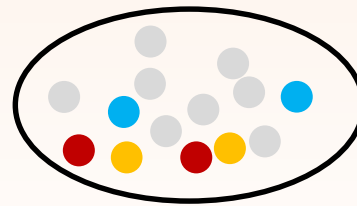
Financial Bank or Riverbank?

... *deposit a cheque at the bank before it closed* ...



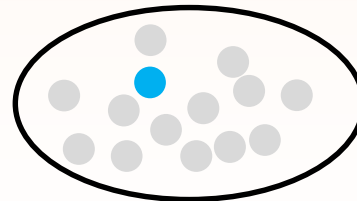
- $overlap(bank\#1, B)$

- 6 overlaps found



- $overlap(bank\#2, B)$

- 1 overlap found



- Decision: select sense 1.

Model Variations

Which dictionary to use? NLTK?

Use only dictionary definitions? Or include example sentences?

Ignore uninformative stopwords (e.g., *the, a, of*)?

Lemmatize when considering matches (*tomatoes* matches *tomato*)?

Exercise

Run the Lesk algorithm using NLTK. Ignore stop words, include examples, count lemma overlap. Consider only the top two senses of bank.

1. I'll deposit the cheque at the **bank**.
2. The **bank** overflowed and water flooded the town.

Yarowsky's Algorithm (1995)

A method based on **bootstrapping**

Steps:

1. Gather a data set with target word to be diambiguated
2. Automatically label a small **seed set** of examples
3. Repeat the following for a while:
 - Train a supervised learning algorithm from the seed set
 - Apply the supervised model to the entire data set
 - Keep the highly confident classification outputs to be the new seed set
4. Use the last model as the final model

Yarowsky's Example

Step 1: Disambiguating *plant*

Sense	Training Examples (Keyword in Context)
?	... company said the <i>plant</i> is still operating
?	Although thousands of <i>plant</i> and animal species
?	... zonal distribution of <i>plant</i> life
?	... to strain microscopic <i>plant</i> life from the ...
?	vinyl chloride monomer <i>plant</i> , which is ...
?	and Golgi apparatus of <i>plant</i> and animal cells
?	... computer disk drive <i>plant</i> located in ...
?	... divide life into <i>plant</i> and animal kingdom
?	... close-up studies of <i>plant</i> life and natural
?	... Nissan car and truck <i>plant</i> in Japan is ...
?	... keep a manufacturing <i>plant</i> profitable without
?	... molecules found in <i>plant</i> and animal tissue
?	... union responses to <i>plant</i> closures
?	... animal rather than <i>plant</i> tissues can be
?	... many dangers to <i>plant</i> and animal life
?	company manufacturing <i>plant</i> is in Orlando ...
?	... growth of aquatic <i>plant</i> life in water ...
?	automated manufacturing <i>plant</i> in Fremont ,
?	... Animal and <i>plant</i> life are delicately
?	discovered at a St. Louis <i>plant</i> manufacturing
?	computer manufacturing <i>plant</i> and adjacent ...
?	... the proliferation of <i>plant</i> and animal life
?

Step 2: Initial Seed Set

Sense A:

- *plant* as in a lifeform

Other data

Sense B:

- *plant* as in a factory

Sense	Training Examples (Keyword in Context)
A	used to strain microscopic <i>plant life</i> from the ...
A	... zonal distribution of <i>plant life</i>
A	close-up studies of <i>plant life</i> and natural ...
A	too rapid growth of aquatic <i>plant life</i> in water ...
A	... the proliferation of <i>plant</i> and animal life ...
A	establishment phase of the <i>plant</i> virus life cycle ...
A	... that divide life into <i>plant</i> and animal kingdom
A	... many dangers to <i>plant</i> and animal life ...
A	mammals . Animal and <i>plant life</i> are delicately
A	beds too salty to support <i>plant life</i> . River ...
A	heavy seas, damage , and <i>plant life</i> growing on ...
A
?	... vinyl chloride monomer <i>plant</i> , which is ...
?	... molecules found in <i>plant</i> and animal tissue
?	... Nissan car and truck <i>plant</i> in Japan is ...
?	... and Golgi apparatus of <i>plant</i> and animal cells ...
?	... union responses to <i>plant</i> closures
?
?
?	... cell types found in the <i>plant</i> kingdom are ...
?	... company said the <i>plant</i> is still operating ...
?	... Although thousands of <i>plant</i> and animal species
?	... animal rather than <i>plant</i> tissues can be ...
?	... computer disk drive <i>plant</i> located in ...
B
B	automated manufacturing <i>plant</i> in Fremont ...
B	... vast manufacturing <i>plant</i> and distribution ...
B	chemical manufacturing <i>plant</i> , producing viscose
B	... keep a manufacturing <i>plant</i> profitable without
B	computer manufacturing <i>plant</i> and adjacent ...
B	discovered at a St. Louis <i>plant manufacturing</i>
B	... copper manufacturing <i>plant</i> found that they
B	copper wire manufacturing <i>plant</i> , for example ...
B	's cement manufacturing <i>plant</i> in Alpena ...
B	polystyrene manufacturing <i>plant</i> at its Dow ...
B	company manufacturing <i>plant</i> is in Orlando ...

Step 3: Train a Classifier

He went with a **decision-list** classifier (we didn't cover this one in class)

Initial decision list for <i>plant</i> (abbreviated)		
LogL	Collocation	Sense
8.10	<i>plant</i> life	⇒ A
7.58	manufacturing <i>plant</i>	⇒ B
7.39	life (within ±2-10 words)	⇒ A
7.20	manufacturing (in ±2-10 words)	⇒ B
6.27	animal (within ±2-10 words)	⇒ A
4.70	equipment (within ±2-10 words)	⇒ B
4.39	employee (within ±2-10 words)	⇒ B
4.30	assembly <i>plant</i>	⇒ B
4.10	<i>plant</i> closure	⇒ B
3.52	<i>plant</i> species	⇒ A
3.48	automate (within ±2-10 words)	⇒ B
3.45	microscopic <i>plant</i>	⇒ A
	...	

Note how new collocations are found for each sense

Step 3: Change Seed Set

Use only the cases where classifier is highly confident

Labeling previously untagged contexts
using the one-sense-per-discourse property

Change in tag	Disc. Numb.	Training Examples (from same discourse)
A → A	724	... the existence of <i>plant</i> and animal life ...
A → A	724	... classified as either <i>plant</i> or animal ...
? → A	724	Although bacterial and <i>plant</i> cells are enclosed
A → A	348	... the life of the <i>plant</i> , producing stem
A → A	348	... an aspect of <i>plant</i> life , for example
? → A	348	... tissues ; because <i>plant</i> egg cells have
? → A	348	photosynthesis, and so <i>plant</i> growth is attuned

Results

96% on binary word sense distinctions

Same result as with supervised methods, but with minimal amounts of annotation effort!