| Lexical Semantics |
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| Chapter 19 |

## Three Perspectives on Meaning

1. Lexical Semantics

- The meanings of individual words

2. Formal Semantics (or Compositional Semantics or Sentential Semantics)

- How those meanings combine to make meanings for individual sentences or utterances

3. Discourse or Pragmatics

- How those meanings combine with each other and with other facts about various kinds of context to make meanings for a text or discourse.
- Dialog or Conversation is often lumped together with Discourse.


## Outline: Computational Lexical Semantics

- Introduction to Lexical Semantics
- Word relations such as Homonymy, Polysemy, Synonymy
- Online resources: WordNet
- Computational Lexical Semantics
- Word Sense Disambiguation
- Supervised
- Semi-supervised
- Word Similarity
- Thesaurus-based
- Distributional


## Preliminaries

- What's a word?
- Definitions we've used over the class: Types, tokens, stems, roots, uninflected forms, etc..
- Lexeme: An entry in a lexicon consisting of a pairing of a form with a single meaning representation
- Lexicon: A collection of lexemes
- Lemma - citation form - uninflected form (used to represent a lexeme). Need to do morphological parsing to get from wordform to lemma (lemmatization)
- Lemma is part-of-speech specific (e.g., table N and V)


## Relationships between word meanings

- Homonymy
- Polysemy
- Synonymy
- Antonymy
- Hypernomy
- Hyponomy
- Meronomy


## Homonymy

- Lexemes that share a form
- Phonological, orthographic or both
- But have unrelated, distinct meanings
- Clear example:
- Bat (wooden stick-like thing) vs
- Bat (flying scary mammal thing)
- Or bank (financial institution) versus bank (riverside)
- Can be homophones, homographs, or both
- Homophones:
- Write and right
- Piece and peace


## Homonymy causes problems for NLP applications

- Text-to-Speech
- Same orthographic form but different phonological form
- Bass vs bass
- Bow vs bow
- Record vs record
- Information retrieval
- Different meanings same orthographic form - QUERY: bat care
- Machine Translation
- Speech recognition


## Polysemy

- The bank is constructed from red brick I withdrew the money from the bank
- Are those the same sense?
- What about river bank?
- What about: The food bank is having a donation drive next week.
- Different senses but some more related than others..
- When two senses are related semantically we call it polysemy (rather than homonymy)


## Metaphor and Metonymy

- Specific types of polysemy
- Metaphor:
- Germany will pull Slovenia out of its economic slump.
- I spent 2 hours on that homework.
- I put money into Google stock.
- Metonymy (use of one aspect of a concept or entity to refer to other aspects of the entity or to the entity itself)
- The White House announced yesterday...
- White House refers to the administration whose office is in the White House
- This chapter talks about part-of-speech tagging
- Bank (building) and bank (financial institution)
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## How do we know when a word has more than one sense?

- ATIS examples
- Which flights serve breakfast?
- Does America West serve Philadelphia?
- The "zeugma" test:
- ?Does United serve breakfast and San Jose?


## Synonyms

- Words that have the same meaning in some or all contexts
- Filbert / hazelnut
- Couch/sofa
- Big / large
- Automobile / car
- Vomit/ throw up
- Water/ $\mathrm{H}_{2} \mathrm{O}$
- Two lexemes are synonyms if they can be successfully substituted for each other in all situations
- If so they have the same propositional meaning


## Synonyms

- But there are few (or no) examples of perfect synonym
- Why should that be?
- Even if many aspects of meaning are identical
- Still may not preserve the acceptability based on notions of politeness, slang, register, genre, etc...
- Example
- Water and $\mathrm{H}_{2} \mathrm{O}$
- Large coke versus *big coke


## Some more terminology

- Lemmas and word forms
- A lexeme is an abstract pairing of meaning and form
- A lemma or citation form is the grammatical form that is used to represent a lexeme
- Carpet is the lemma for carpets
- Corpus is the lemma for corpora
- Specific surface forms carpets, sung, corpora are called wordforms
- The lemma bank has two senses:
- Instead, a bank can hold the investments in..
- But as agriculture burgeons on the east bank, the river will shrink even more
- A sense is a discrete representation of one aspect of the meaning of a word


## Synonymy is a relation between senses rather than words

- Consider the words big and large
- Are they synonyms?
- How big is that plane?
- Would I be flying on a large or small plane?
- How about here:
- Miss Nelson, for instance, became a kind of big sister to Benjamin.
- ?Miss Nelson, for instance, became a kind of large sister to Benjamin.
- Why?
- Big has a sense that means being older, or grown up
- Large lacks this sense


## Hyponym

- One sense is a hyponym of another if the first sense is more specific, denoting a subclass of the other
- Car is a hyponym of vehicle
- Dog is a hyponym of animal
- Mango is a hyponym of fruit
- Conversely
- Vehicle is a hypernym/superordinate of car
- Animal is a hypernym of dog
- Fruit is a hypernym of mango

| Superordinate | Vehicle | Fruit | Furniture | mammal |
| :--- | :--- | :--- | :--- | :--- |
| Hyponym | Car | Mango | Chair | Dog |

## Hyponymy more formally

- Extensional:
- The class denoted by the superordinate extensionally includes the class denoted by the hyponym
- Entailment
- A sense $A$ is a hyponym of sense $B$ if being an $A$ entails being a $B$
- Hyponymy is usually transitive
- (A hypo B and B hypo C entails A hypo C)


## II. Wordnet

- A hierarchically organized lexical database
- On-line thesaurus + aspects of a dictionary

| Category | Unique Forms |
| :--- | :--- |
| Noun | 117,097 |
| Verb | 11,488 |
| Adjective | 22,141 |
| Adverb | 4,601 |

## Format of WordNet Entries

- The noun bass has 8 senses in wordnet:
- $\mathrm{S}:(\mathrm{n})$ bass (the lowest part of the musical range)
- $\underline{\text { : }}(\mathrm{n})$ bass, bass part (the lowest part in polyphonic music)
- S: ( n ) bass, basso (an adult male singer with the lowest voice)
- S: ( $n$ ) sea bass, bass (the lean flesh of a saltwater fish of the family Serranidae)
- S: (n) freshwater bass, bass (any of various North American freshwater fish with lean flesh (especially of the genus Micropterus))
- S: ( n ) bass, bass voice, basso (the lowest adult male singing voice)
- $\mathrm{S}:(\mathrm{n})$ bass (the member with the lowest range of a family of musical instruments)
- S: (n) bass (nontechnical name for any of numerous edible marine and freshwater spiny-finned fishes)
- And 1 Adjective Sense:
- $\underline{\text { S: }}$ (adj) bass, deep (having or denoting a low vocal or instrumental range) "a deep voice"; "a bass voice is lower than a baritone voice"; "a bass clarinet"

WordNet Noun Relations

| Relation <br> Hypernym <br> Hyponym <br> Member Meronym <br> Has-Instance <br> Instance <br> Member Holonym <br> Part Meronym <br> Part Holonym <br> Antonym | Also called <br> Superordinate <br> Subordinate <br> Has-Member <br>  <br>  <br> Member-Of <br> Has-Part <br> Part-Of | Definition <br> From concepts to superordinates <br> From concepts to subtypes <br> From groups to their members <br> From concepts to instances of the concept <br> From instances to their concepts <br> From members to their groups <br> From wholes to parts <br> From parts to wholes <br> Opposites | Example Sreakfasi $^{1} \rightarrow$ meal $^{1}$ meal $^{1} \rightarrow$ lumh $^{1}$ facully $^{2} \rightarrow$ Professor $^{1}$ composer $^{1} \rightarrow$ Bach $^{1}$ Austen copilor $^{1} \rightarrow$ author $^{1}$ tablew $^{2} \rightarrow$ leg $^{3}$ course $^{7} \rightarrow$ meal leader $^{1} \rightarrow$ follower $^{1}$ |
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WordNet Hierarchies
sense 3
(an adult male singer with the lovest voice)
#> singer, vocalist, vocalizer, vocaliser
    => musician, instrumentalist, player
        -> performor, performing artist
        Merformor, porfo
            >> person, individual, somoone..
            porson, individual,
            #
                ->objoct, physical object
                #> physical entity
            -> causal agent, cause, causal agency
            # physical ent
sense ?
lol
->> musical instrument, ingtrument
    #> dovice
        -> instrumentality, inst,
        #- whole, artefact
            vhole, unit mbyical object
            -> physical entity
                physical entity
```


## How is "sense" defined in WordNet?

- The set of near-synonyms for a WordNet sense is called a synset (synonym set); it's their version of a sense or a concept.
- Example: chump as a noun to mean
- 'a person who is gullible and easy to take advantage of'
- chump\#1, fool\#2, gull\#1, mark\#9, patsy\#1, fall guy\#1, sucker\#1, soft touch\#1, mug\#2 (a person who is gullible and easy to take advantage of)
- Each of these senses share this same gloss
- Thus, for WordNet, the meaning of this sense of chump is this list.

Word Sense Disambiguation (WSD)

- Given
- A word in context,
- A fixed inventory of potential word senses
- Decide which sense of the word this is
- English-to-Spanish MT
- Inventory is the set of Spanish translations
- Speech Synthesis
- Inventory is homographs with different pronunciations like bass and bow
- Automatic indexing of medical articles
- MeSH (Medical Subject Headings) thesaurus entries

Approaches

- Supervised
- Semi-supervised
- Unsupervised
- Dictionary-based techniques
- Selectional association
- Lightly supervised
- Bootstrapping
- Preferred Selectional Association

| Approaches |
| :---: |
| - Supervised |
| • Semi-supervised |
| - Unsupervised |
| $\quad$ • Dictionary-based techniques |
| - Selectional association |
| • Bootstrapping |
| • Preferred Selectional Association |

## Two variants of WSD task

- Lexical Sample task
- Small pre-selected set of target words
- And inventory of senses for each word
- All-words task
- Every word in an entire text
- A lexicon with senses for each word
- Sort-of like part-of-speech tagging
- Except each lemma has its own tagset


## Supervised Machine Learning Approaches

- Supervised machine learning approach:
- A training corpus of?
- Used to train a classifier that can tag words in text
- Just as in part-of-speech tagging, statistical MT.
- Summary of what we need:
- The tag set ("sense inventory")
- The training corpus
- A set of features extracted from the training corpus
- A classifier


## Supervised WSD 1: WSD Tag

- What's a tag?


## WordNet Bass

- The noun "bass" has 8 senses in WordNet
- $\underline{\text { S: }}$ : $n$ ) bass\#1 (the lowest part of the musical range)
- $\underline{S}:(n)$ bass\#2, bass part\#1 (the lowest part in polyphonic music)
- $\underline{\text { S: }}$ ( $n$ ) bass\#3, basso\#1 (an adult male singer with the lowest voice)
- S: ( n ) sea bass\#1, bass\#4 (the lean flesh of a saltwater fish of the family Serranidae)
- S: (n) freshwater bass\#1, bass\#5 (any of various North American freshwater fish with lean flesh (especially of the genus Micropterus))
- $\underline{S}$ : (n) bass\#6, bass voice\#1, basso\#2 (the lowest adult male singing voice)
- $\underline{\text { S: }}$ ( $n$ ) bass\#7 (the member with the lowest range of a family of musical instruments)
- S: ( n ) bass\#8 (nontechnical name for any of numerous edible marine and freshwater spiny-finned fishes)

Inventory of sense tags for bass
Supervised WSD 2: Get a corpus

- Lexical sample task:
- Line-hard-serve corpus -4000 examples of each
- Interestcorpus -2369 sense-tagged examples
- All words:
- Semantic concordance: a corpus in which each open-class word is labeled with a sense from a specific dictionary/thesaurus.
- SemCor: 234,000 words from Brown Corpus, manually tagged with WordNet senses
- SENSEVAL-3 competition corpora -2081 tagged word tokens


## Supervised WSD 3: Extract feature vectors

- Weaver (1955)
- If one examines the words in a book, one at a time as through an opaque mask with a hole in it one word wide, then it is obviously impossible to determine, one at a time, the meaning of the words. [...] But if one lengthens the slit in the opaque mask, until one can see not only the central word in question but also say N words on either side, then if N is large enough one can unambiguously decide the meaning of the central word. [...] The practical question is : "What minimum value of N will, at least in a tolerable fraction of cases, lead to the correct choice of meaning for the central word?"
- washing dishes .
- simple dishes including
- convenient dishes to
- of dishes and
- free bass with
- pound bass of
- and bass player
- his bass while
- We need more good teachers -right now, there are only a half a dozen who can play the free bass with ease.
- Though still a far cry from the lake's record 52-pound bass of a decade ago, "you could fillet these fish again, and that made people very, very happy." Mr. Paulson says.
- An electric guitar and bass player stand off to one side, not really part of the scene, just as a sort of nod to gringo expectations again.
- Lowe caught his bass while fishing with pro Bill Lee of Killeen, Texas, who is currently in 144th place with two bass weighing 2-09.

Two kinds of features in the vectors

- Collocational features and bag-of-words features
- Collocational
- Features about words at specific positions near target word
- Often limited to just word identity and POS
- Bag-of-words
- Features about words that occur anywhere in the window (regardless of position)
- Typically limited to frequency counts
- "In our house, everybody has a career and none of them includes washing dishes," he says.
- In her tiny kitchen at home, Ms. Chen works efficiently, stir-frying several simple dishes, including braised pig's ears and chcken livers with green peppers.
- Post quick and convenient dishes to fix when your in a hurry.
- Japanese cuisine offers a great variety of dishes and regional specialties


## Feature Vectors

- A simple representation for each observation (each instance of a target word)
- Vectors of sets of feature/value pairs
- I.e. files of comma-separated values
- These vectors should represent the window of words around the target

> How big should that window be?

## Examples

Example text (WSJ)

- An electric guitar and bass player stand off to one side not really part of the scene, just as a sort of nod to gringo expectations perhaps
- Assume a window of +/-2 from the target


## Examples

Example text (WSJ)

- An electric guitar and bass player stand off to one side not really part of the scene, just as a sort of nod to gringo expectations perhaps
- Assume a window of $+/-2$ from the target


## Collocational

- Position-specific information about the words in the window
- guitar and bass player stand
- [guitar, NN, and, CC, player, NN, stand, VB]
- Wordn-2,POSn-2,wordn-1,POSn1,Wordn+1POSn+1...
- In other words, a vector consisting of
- [position n word, position n part-of-speech...]


## Bag-of-words

- Information about the words that occur within the window.
- First derive a set of terms to place in the vector.
- Then note how often each of those terms occurs in a given window.


## Co-Occurrence Example

- Assume we've settled on a possible vocabulary of 12 words that includes guitar and player but not and and stand
- guitar and bass player stand
- [0,0,0,1,0,0,0,0,0,1,0,0]
- Which are the counts of words predefined as e.g.,
- [fish,fishing,viol, guitar, double,cello...


## Classifiers

- Once we cast the WSD problem as a classification problem, then all sorts of techniques are possible
- Naïve Bayes (the easiest thing to try first)
- Decision lists
- Decision trees
- Neural nets
- Support vector machines
- Nearest neighbor methods...


## WSD Evaluations and Baselines

- In vivo (end-to-end, extrinsic, task-based) versus in vitro (intrinsic as if a stand-alone system) evaluation
- In vitro evaluation is most common now
- Exact match accuracy
- \% of words tagged identically with manual sense tags
- Usually evaluate using held-out data from same labeled corpus - Problems?
- Why do we do it anyhow?
- Baselines
- Most frequent sense
- The Lesk algorithm (choose the sense whose dictionary gloss or definition shares the most words with the target word's neighborhood.


## Most Frequent Sense

- WordNet senses are order in frequency order
- So "most frequent sense" in WordNet = "take the first sense"

Gloss
Freq Synset
338 plant ${ }^{1}$, works, industrial plant buildings for carrying on industrial labor
207 plant ${ }^{2}$, flora, plant life
2 plant ${ }^{3}$ plant ${ }^{4}$ something planted secretly for discovery by another an actor situated in the audience whose acting is rehearsed but an actor situated in the audience whens
seems spontaneous to the audience

## Ceiling

- Human-inter-annotator agreement
- Compare annotations of two humans
- On same data
- Given same tagging guidelines
- Human agreements on all-words corpora with WordNet style senses
- 75\%-80\%

