# **CMPT-413: Computational Linguistics**

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### **Lexical Semantics**

- So far, we have listed words in our lexicon or vocabulary assuming a single meaning per word:
  - Consider n-grams  $P(w_i \mid w_{i-2}, w_{i-1}) = P(Bank \mid on, Commerce)$  or prepositional phrase attachment if p=on and n2=bank then change N to V
- Consider . . . withdraw twenty dollars on the bank (correct = V) vs. . . . withdraw the troops on the bank (correct = N)
- The same word bank means two different things but we cannot distinguish between them using the traditional definition of word.

#### **Lexical Semantics**

- To deal with this issue, we combine the spelling or pronunciation of a word and the meaning.
   In the lexicon we now store lexemes instead of words. A lexeme pairs a particular spelling or pronunciation with a particular meaning.
- The meaning part of a lexeme is called a sense. For CL, our interest is in relations between lexemes or disambiguating different senses of a word. word: bank → lexeme: bank<sup>1</sup> OR word: bank → lexeme: bank<sup>2</sup>
- Note that meanings are often not definitions, but often are simple listings of compatible lexemes.
   cf. dictionary defns: red, n. the color of blood or ruby; blood, n. red liquid circulating in animals

### Homonyms

- Homonyms: words that have the same form but different meanings
  - 1. Instead, the chemical plant was found in violation of several environmental laws
  - 2. Stanley formed an expedition to find a rare plant found along the Amazon river
- Same *orthographic* form: *plant* but two senses: **plant**<sup>1</sup> and **plant**<sup>2</sup>

### Homonyms

- Text vs. speech: fly-casting for bass vs. rhythmic bass chords
   These cases are homonyms in text, but not in speech. Referred to as homographs
- Speech vs. text: would vs. wood
   These cases are not homonyms in text, but easily confused in speech.
   Referred to as homophones
- Note that this problem in some cases can be solved using part of speech tagging
  - Can you think of a case which cannot be solved using POS tagging?

## **Applications**

- Spelling correction: homophones: weather vs. whether
- Speech recognition: homophones: *to*, *two*, *too*. Also homonyms (see *n*-gram e.g.)
- Text to speech: homographs: bass vs. bass
- Information retrieval: homonyms: *latex*

## Polysemy

- Consider the homonym:  $bank \rightarrow commercial bank^1$  vs. river  $bank^2$
- Now consider
  - 1. A PCFG can be trained using derivation trees from a tree bank annotated by human experts
- Is this a new sense of bank?

### Polysemy

- Senses can be derived from a particular lexeme. This process is known as polysemy
  - In previous case we would say that the use of *bank* is a sense derived from commercial **bank**<sup>1</sup>
- In some cases, splitting into different lexemes has other supporting evidence: **bank**<sup>1</sup> has Italian origin vs. **bank**<sup>2</sup> has Scandinavian origin
  - 1. A PCFG can be trained using a bank of derivation trees called a tree-bank annotated by human experts
- How can we tell between homonyms and polysemous uses of a word?

## Zeugma

- Consider the case for a verb like serve
  - 1. Does United serve breakfast?
  - 2. Does United serve Philadelphia?
  - 3. Does United serve breakfast and dinner?
  - 4. #Does United serve breakfast and Philadelphia?

### Word Sense Disambiguation

- Consider a noun like bank
  - 1. How many senses does it have?
  - 2. How are these senses related?
  - 3. How can they be reliably distinguished?
- For NLP software, among these three questions, typically at runtime we need to automatically find the answer to the last question: given a word in context, map it to the correct lexeme: word-sense disambiguation

#### Word Sense Disambiguation: training data

```
training_VBG new_JJ Ukrainian_JJ
                                       plant(1)
                                                   _NN operators_NNS to_TO replace_V
who_WP are_VBP leaving_VBG the_DT
                                       plant(1)
                                                   s_NNS in_IN Ukraine_NNP and_CC im
CC safety_NN procedures_NNS at_IN
                                                   s_NNS in_IN both_DT countries_NNS
                                       plant(1)
t_IN the_DT Orange_NNP County_NNP
                                       plant(1)
                                                   _NN ._.
Z closing_VBG three_CD missile_NN
                                       plant(1)
                                                   s_NNS in_IN southern_JJ Californi
_IN the_DT whole_JJ Chernobyl_NNP
                                       plant(1)
                                                   _NN in_IN 1991_CD ,_, five_CD yea
IN a_DT hill_NN ,_, gardeners_NNS
                                       plant(2)
                                                   _NN begonias_NNS ,_, making_VBG f
$_$ 200_CD million_CD printing_NN
                                       plant(1)
                                                   _NN in_IN Brooklyn_NNP ,_, Ohio_N
of_IN incompletely_JJ oxidated_JJ
                                       plant(2)
                                                   _NN and_CC animal_NN sediment_NN
                                                   _NN ._. ','
whenever_WRB you_PRP eat_VBP a_DT
                                       plant(2)
n_IN return_NN for_IN a_DT new_JJ
                                       plant(1)
                                                   _NN near_IN Tuscaloosa_NNP ._.
                                       plant(1)
T carmaker_NN could_MD finance_VB
                                                   _NN construction_NN with_IN the_D
                                                   _NN near_IN Tuscaloosa_NNP ._.
n_IN return_NN for_IN a_DT new_JJ
                                       plant(1)
```

### Word Sense Disambiguation: learning

Many different learning methods: let's consider one, Transformation
 Based Learning

• Let rule condition

$$r \leftarrow W_{-1} = gardeners, W_{+1} = begonias, W_{+window} = floral$$

• If r then change from **plant**<sup>1</sup> (manufacturing plant) to **plant**<sup>2</sup> (living plant)

## Synonyms

- Synonyms: Different lexemes with the same meaning
  - 1. How big/large is that plane?
  - 2. Would I be flying on a big/large or small plane?
- Synonyms clash with polysemous meanings
  - 1. Seema is my big sister
  - 2. #Seema is my large sister

#### WordNet

- WordNet is an electronic database of word relationships, handcrafted from scratch by researchers at Princeton University (George Miller, Christine Fellbaum, et al.)
- WordNet contains 3 databases: for verbs, nouns and one for adjectives and adverbs

Category	Unique Forms	Number of Senses
Noun	94474	116317
Verb	10319	22066
Adjective	20170	29881
Adverb	4546	5677

#### WordNet

- Ask the question: how many senses per noun or verb? The distribution of senses follows Zipf's (2nd) Law.
- WordNet provides multiple lexeme entries for each word and for each part of speech,
  - e.g. plant as noun has 3 senses; plant as verb has 2 senses
- WordNet also provides domain-independent lexical relations such as IS-A, HasMember, MemberOf, . . .

## WordNet: noun relations

Relation	Definition	Example
Hypernym	this is a kind of	breakfast → meal
Hyponym	this has a specific instance	$meal \rightarrow lunch$
Has-Member	this has a member	$faculty \rightarrow professor$
Member-Of	this is member of a group	copilot  ightarrow crew
Has-Part	this has a part	table  ightarrow leg
Part-Of	this is part of	course  o meal
Antonym	this is an opposite of	$leader \rightarrow follower$

## WordNet: verb relations

Relation	Definition	Example
Hypernym	this event is a kind of	$fly \rightarrow travel$
Tropynym	this event has a subtype	walk $\rightarrow$ stroll
Entails	this event entails	snore $ ightarrow$ sleep
Antonym	this event is opposite of	$increase \rightarrow decrease$

### WordNet: example from ver1.7.1

```
Sense1: Canada

⇒North American country,North American nation
⇒country, state, land
⇒administrative district,administrative division,territorial division
⇒district, territory
⇒region
⇒location
⇒entity, physical thing
```

### WordNet: example from ver1.7.1

```
Sense 3: Vancouver
   ⇒city, metropolis, urban center
       ⇒municipality
           ⇒urban area
               ⇒geographical area
                  ⇒region
                      ⇒location
                          ⇒entity, physical thing
           ⇒administrative district, territorial division
               ⇒district, territory
                  ⇒region
                      ⇒location
                          ⇒entity, physical thing
   ⇒port
       ⇒geographic point
           ⇒point
               ⇒location
                  ⇒entity, physical thing
```

#### WordNet

- A synset in WordNet is a list of synonyms (interchangeable words)
- { chump, fish, fool, gull, mark, patsy, fall guy, sucker, schlemiel, shlemiel, soft touch, mug }
- How can we use this information like synsets, hypernyms, etc. from WordNet to benefit NLP applications?
- Consider one example: PP attachment, words plus word classes extracted from the hypernym hierarchy increase accuracy from 84% to 88% (Stetina and Nagao, 1998)

#### WordNet

- Another example of WordNet used in NLP applications: selectional restrictions
- We have considered subcategorization:
   VP-with-NP-complement → V(eat) NP "eat six bowls of rice"
   But not selectional restrictions of the verb itself: "eat tomorrow"
   Consider what do you want to eat tomorrow
- We can use the synset { food, nutrient } to describe the NP argument of eat then the 60K lexemes under these nodes in the WordNet hierarchy will be acceptable. (however, what about "eat my shorts")
  - → several other applications have been explored