Empirical Methods in Natural Language Processing Lecture 11 Word Sense Disambiguation

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Word Senses

- Some words have multiple meanings
- This is called **Polysemy**
- Example: bank
 - financial institution: I put my money in the bank.
 - river shore: He rested at the bank of the river.
- How could a computer tell these senses apart?



Homonym

- Sometimes two completely different words are spelled the same
- This is called a **Homonym**
- Example: *can*
 - modal verb: You can do it!
 - container: She bought a can of soda.
- Distinction between Polysemy and Homonymy not always clear



How many senses?

- How many senses does the word interest have?
 - She pays 3% interest on the loan.
 - He showed a lot of interest in the painting.
 - Microsoft purchased a controlling interest in Google.
 - It is in the national interest to invade the Bahamas.
 - I only have your best interest in mind.
 - Playing chess is one of my interests.
 - Business interests lobbied for the legislation.
- Are these seven different senses? Four? Three?



Wordnet

- One way to define senses is to look them up in Wordnet, a hierarchical database of senses
- According to Wordnet, interest has 7 senses:
 - Sense 1: a sense of concern with and curiosity about someone or something,
 Synonym: involvement
 - Sense 2: the power of attracting or holding one's interest (because it is unusual or exciting etc.), Synonym: interestingness
 - Sense 3: a reason for wanting something done, Synonym: sake
 - Sense 4: a fixed charge for borrowing money; usually a percentage of the amount borrowed
 - Sense 5: a diversion that occupies one's time and thoughts (usually



- pleasantly), Synonyms: pastime, pursuit
- Sense 6: a right or legal share of something; a financial involvement with something, Synonym: stake
- Sense 7: (usually plural) a social group whose members control some field of activity and who have common aims, Synonym: interest group
- Organization of Wordnet
 - Wordnet groups words into synsets.
 - polysemous words are part of multiple synsets
 - synsets are organized into a hierarchical structure of is-a relationships, e.g.
 a dog is-a pet, pet is-a animal
- Is Wordnet too fine grained?



Different sense = different translation

- Another way to define senses: if occurrences of the word have different translations, these indicate different sense
- Example *interest* translated into German
 - Zins: financial charge paid for load (Wordnet sense 4)
 - Anteil: stake in a company (Wordnet sense 6)
 - *Interesse*: all other senses

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Languages differ

- Foreign language may make finer distinctions
- Translations of *river* into French
 - *fleuve*: river that flows into the sea
 - rivière: smaller river
- English may make finer distinctions than a foreign language
- Translations of German *Sicherheit* into English
 - security
 - safety
 - confidence



One last word on senses

- A lot of research in word sense disambiguation is focused on polysemous words with clearly distinct meanings, e.g. bank, plant, bat, ...
- Often meanings are close and hard to tell apart, e.g. *area*, *field*, *domain*, *part*, *member*, ...
 - She is a part of the team.
 - She is a member of the team.
 - The wheel is a part of the car.
 - * The wheel is a member of the car.



Word sense disambiguation (WSD)

- For many applications, we would like to disambiguate senses
 - we may be only interested in one sense
 - searching for chemical plant on the web, we do not want to know about chemicals in bananas
- Task: Given a polysemous word, find the sense in a given *context*
- Popular topic, data driven methods perform well

WSD as supervised learning problem

- Words can be labeled with their senses
 - She pays 3% interest/INTEREST-MONEY on the loan.
 - He showed a lot of interest/INTEREST-CURIOSITY in the painting.
- Similar to *tagging*
 - given a corpus tagged with senses
 - define features that indicate one sense over another
 - learn a model that predicts the correct sense given the features
- We can apply similar supervised learning methods
 - Naive Bayes, related to HMM
 - Transformation-based learning
 - Maximum entropy learning

Simple features

- Directly neighboring words
 - plant life
 - manufacturing plant
 - assembly plant
 - plant closure
 - plant species
- Any content words in a 10 word window (also larger windows)
 - animal
 - equipment
 - employee
 - automatic



More features

- Syntactically related words
- Syntactic role in sense
- Topic of the text
- Part-of-speech tag, surrounding part-of-speech tags

Training data for supervised WSD

- SENSEVAL competition
 - bi-annual competition on WSD
 - provides annotated corpora in many languages
- Pseudo-words
 - create artificial corpus by artificially conflate words
 - example: replace all occurrences of banana and door with banana-door
- Multi-lingual parallel corpora
 - translated texts aligned at the sentence level
 - translation indicates sense

Naive Bayes

- ullet We want to predict the sense S given a set of features F
- First, apply the Bayes rule

$$\operatorname{argmax}_{S} p(S|F) = \operatorname{argmax}_{S} p(F|S) p(F) \tag{1}$$

• Then, decompose p(F) by assuming all features are independent (that's *naive!*)

$$p(F) = \prod_{f_i \in F} p(f_i|S) \tag{2}$$

• The prior p(S) and the conditional posterior probabilities $p(f_i|S)$ can be learned by maximum likelihood estimation

Decision list

- Yarowsky [1994] uses a decision list for WSD
 - two senses per word
 - rules of the form: collocation \rightarrow sense
 - example: manufacturing plant → PLANT-FACTORY
 - rules are ordered, most reliable rules first
 - when classifying a test example, step through the list, make decision on first rule that applies
- Learning: rules are ordered by

$$\log\left(\frac{p(sense_A|collocation_i)}{p(sense_B|collocation_i)}\right) \tag{3}$$

Smoothing is important

Bootstrapping

- Yarowsky [1995] presents bootstrapping method
 - 1. label a few examples
 - 2. learn a decision list
 - 3. apply decision list to unlabeled examples, thus labeling them
 - 4. add newly labeled examples to training set
 - 5. go to step 2, until no more examples can be labeled
- Initial starting point could also be
 - a short decision list
 - words from dictionary definition



One sense per discourse

- Rules encode the principle:
 One sense per collocation
- Bootstrapping method also uses important principle:
 One sense per discourse
 - in one discourse only one sense of a polysemous word appears
 - text talks either about PLANT-FACTORY or PLANT-LIVING
- Improved bootstrapping method
 - after labeling examples, one sense per discourse principle is enforced
 - all examples in one document are labeled with the same sense
 - or, examples that are not in the majority sense are un-labeled