


CS4740 Natural Language Processing

- Last classes

- Intro to lexical semantics
- Lexical semantic resources: WordNet

- Next

-  – Word sense disambiguation
 - » Dictionary-based approaches
 - » Supervised machine learning methods
 - » WSD evaluation
 - » Weakly supervised methods

Word sense disambiguation

- Given a *fixed* set of senses associated with a lexical item (usually an orthographic form rather than a lexeme),
- determine which sense applies to a particular instance of the lexical item in context (i.e., in running text)

Marseille found a smelly **bass**
along the **bank** of the Seine.

Example

Marseille is annoying when he **begs** for his dinner.

Verb


- S: (v) **beg**#1, implore#1, pray#2 (call upon in supplication; entreat)
- S: (v) solicit#1, **beg**#2, tap#12 (make a solicitation or entreaty for something; request urgently or persistently)
- S: (v) **beg**#3 (ask to obtain free)
- S: (v) **beg**#4 (dodge, avoid answering, or take for granted)

Example

Marseille is annoying when he **begs** for his dinner.

- S: (v) **beg**#1, implore#1, pray#2 (call upon in supplication; entreat) *"I beg you to stop!"*
- S: (v) solicit#1, **beg**#2, tap#12 (make a solicitation or entreaty for something; request urgently or persistently) *"Henry IV solicited the Pope for a divorce"; "My neighbor keeps soliciting money for different charities"*
- S: (v) **beg**#3 (ask to obtain free) *"beg money and food"*
- S: (v) **beg**#4 (dodge, avoid answering, or take for granted) *"beg the question"; "beg the point in the discussion"*

Word sense disambiguation

- Two fundamental approaches
 - WSD occurs during semantic analysis as a side-effect of the elimination of ill-formed semantic representations
 -  – Stand-alone approach
 - » WSD is performed independent of, and prior to, compositional semantic analysis
 - » Makes minimal assumptions about what information will be available from other NLP processes
 - » Applicable in large-scale practical applications

Dictionary-based approaches

- Rely on machine readable dictionaries
- Initial implementation of this kind of approach is due to Michael Lesk (1986)
 - Given a word W to be disambiguated in context C
 - » Retrieve all of the sense definitions for W , S_W , from the MRD
 - » Compare each s in S_W to D_C --- all of the dictionary definitions of all words in C
 - » Select the sense s with the largest content-word overlap with D_C

Example

- $W = \textit{cone}$
 - $C = \{ \textit{pine} \}$ ***pine cone***
 - Sense definitions from MRD
 - pine* 1 kind of evergreen tree with needle-shaped leaves
 - 2 waste away through sorrow or illness
 - cone* 1 solid body which narrows to a point
 - 2 something of this shape whether solid or hollow
 - 3 fruit of certain evergreen trees
- } D_C
- } S_W
- Accuracy of 50-70% on short samples of text from *Pride and Prejudice* and an AP newswire article.

CS4740 Natural Language Processing

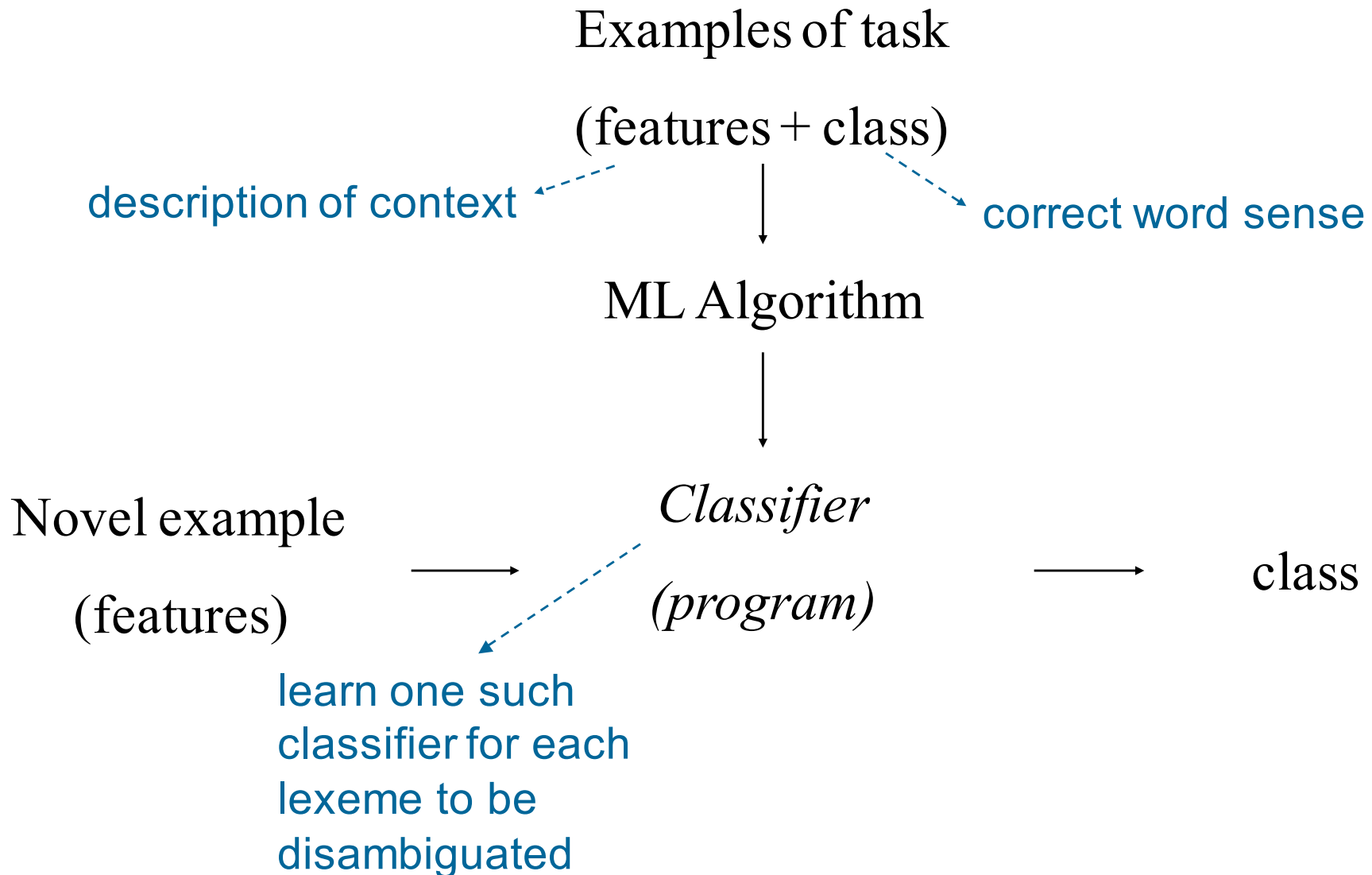
- Last classes
 - Intro to lexical semantics
 - Lexical semantic resources: WordNet
- Next
 - Word sense disambiguation
 - » Dictionary-based approaches
 - » Supervised machine learning methods
 - » WSD evaluation
 - » Weakly supervised methods



Machine learning approaches

- Machine learning paradigms for WSD
 - *Supervised inductive* learning
 - classification
 - Bootstrapping
 - Unsupervised
- Emphasis is on acquiring the knowledge needed for the task from data, rather than from human analysts (e.g., via a set of rules) or from a static algorithm (e.g., Lesk approach)

Supervised ML framework



Running example

*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.*



- 1 Fish sense
- 2 Musical sense
- 3 ...

Feature vector representation

- W.r.t. the **target**, i.e. the word to be disambiguated
- Describe **context** : portion of the surrounding text
 - Select a “window” size
 - Extract features from the context (and possibly the target)
 - » Attribute-value pairs
 - » Values can be numeric, boolean, categorical, ...

What features to use?

Marseille is annoying when he **begs** for his dinner.

- S: (v) **beg**#1, implore#1, pray#2 (call upon in supplication; entreat) *"I beg you to stop!"*
- S: (v) solicit#1, **beg**#2, tap#12 (make a solicitation or entreaty for something; request urgently or persistently) *"Henry IV solicited the Pope for a divorce"; "My neighbor keeps soliciting money for different charities"*
- S: (v) **beg**#3 (ask to obtain free) *"beg money and food"*
- S: (v) **beg**#4 (dodge, avoid answering, or take for granted) *"beg the question"; "beg the point in the discussion"*

What features to use?

*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.*

Collocational features

- Encode information about the lexical inhabitants of *specific* positions located to the left or right of the target word.
 - E.g. the word, its root form, its part-of-speech
 - *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.*

| | | | | | | | |
|------------------|-----------------|------------------|-----------------|------------------|-----------------|------------------|-----------------|
| <u>pre2-word</u> | <u>pre2-pos</u> | <u>pre1-word</u> | <u>pre1-pos</u> | <u>fol1-word</u> | <u>fol1-pos</u> | <u>fol2-word</u> | <u>fol2-pos</u> |
| guitar | NN1 | and | CJC | player | NN1 | stand | VVB |

Co-occurrence features

- Encode information about neighboring words, ignoring exact positions.
 - **Attributes:** words highly associated with exactly one of the senses
 - **Values:** number of times the word occurs in a region surrounding the target word
 - Select a small number of frequently used content words for use as **attributes (features)**
 - » n most frequent content words from a collection of *bass* sentences drawn from the WSJ: *fishing, big, sound, player, fly, rod, pound, double, runs, playing, guitar, band*
 - » window of size 10

| | | | | | | | | | | |
|-----------------|-------------|---------------|----------------|-------------|-------------|---------------|----------------|-----|----------------|--------------|
| <u>fishing?</u> | <u>big?</u> | <u>sound?</u> | <u>player?</u> | <u>fly?</u> | <u>rod?</u> | <u>pound?</u> | <u>double?</u> | ... | <u>guitar?</u> | <u>band?</u> |
| 0 | 0 | 0 | 1 | 0 | 0 | 0 | 0 | | 1 | 0 |

Labeled training example

- *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.*

| <u>pre2-word</u> | <u>pre2-pos</u> | <u>pre1-word</u> | <u>pre1-pos</u> | <u>fol1-word</u> | <u>fol1-pos</u> | <u>fol2-word</u> | <u>fol2-pos</u> | | | |
|------------------|-----------------|------------------|-----------------|------------------|-----------------|------------------|-----------------|-----|----------------|--------------|
| guitar | NN1 | and | CJC | player | NN1 | stand | VVB | | | |
| <u>fishing?</u> | <u>big?</u> | <u>sound?</u> | <u>player?</u> | <u>fly?</u> | <u>rod?</u> | <u>pound?</u> | <u>double?</u> | ... | <u>guitar?</u> | <u>band?</u> |
| 0 | 0 | 0 | 1 | 0 | 0 | 0 | 0 | | 1 | 0 |

: *music*

guitar, NN1, and, CJC, player, NN1, stand, VVB, 0, 0, 0, 1, 0, ..., 1, 0 : music