

# Natural Language Processing

## Part 4: lexical semantics

# Lexical semantics

- A **lexicon** generally has a highly structured form
  - It stores the meanings and uses of each word
  - It encodes the relations between words and meanings
- A **lexeme** is the minimal unit represented in the lexicon
  - It pairs a stem (the orthographic/phonological form chosen to represent words) with a symbolic form for meaning representation (**sense**)
- A **dictionary** is a kind of lexicon where meanings are expressed through definitions and examples

**son** noun

a boy or man in relation to either or both of his parents.

- a male offspring of an animal.
- a male descendant : *the sons of Adam*.
- ( the Son) (in Christian belief) the second person of the Trinity; Christ.
- a man considered in relation to his native country or area : *one of Nevada's most famous sons*.
- a man regarded as the product of a particular person, influence, or environment : *sons of the French Revolution*.
- (also my son) used by an elder person as a form of address for a boy or young man : “*You're on private land, son.*”

# Lexicons & dictionaries

- Definitions in dictionaries exploit words and they may be circular (a word definition uses words whose definitions exploit that word)

**right** adj.

1. of, relating to, situated on, or being the side of the body which is away from the side on which the heart is mostly located
2. located nearer to the **right** hand than to the left
3. done with the **right** hand
- .....

- The paradox is that the dictionary elements are not direct definitions
  - They are descriptions of the lexemes made up of other lexemes assuming that the user has enough information on these other terms!
  - This approach would fail without the assumption that the user has already enough a priori knowledge deriving from the real world
  - However the descriptions provide a great amount of information on the relationships among the words allowing to perform semantic inferences

# Relationships & senses

- Several kinds of relationships can be defined between lexemes and senses (some of them are important for automatic processing)
  - **Homonymy**  
It is a relation between words that have the same form (and the same PoS) but unrelated meanings
    - e.g. *bank* (the financial institution, the river bank)
    - It causes ambiguities for the interpretation of a sentence since it defines a set of different lexemes with the same orthographic form (*bank*<sup>1</sup>, *bank*<sup>2</sup>,...)
    - Related properties are homophony (same pronunciation but different orthography, e.g. *be-bee*) and homography (same orthography but different pronunciation *pésca/pèsca*)
  - **Polysemy**  
It happens when a lexeme has more related meanings
    - It depends on the word etymology (unrelated meanings usually have a different origin) - e.g. *bank/data bank/blood bank*

# Polysemy / Synonymy

For polysemous lexemes we need to manage all the meanings

- We should define a method to determine the meanings (their number and semantics) and if they are really distinct (by experts in lexicography)
- We need to describe the eventual correlations among the meanings
- We need to define how the meanings can be distinguished in order to attach the correct meaning to a word in a given context (word sense disambiguation)

## ▫ **Synonymy**

It is a relationship between two distinct lexemes with the same meaning (i.e. they can be substituted for one another in a given context without changing its meaning and correctness) – e.g. I received a gift/present

- The substitutability may not be valid for any context due to small semantic differences (e.g. *price/fare of a service – the bus fare/the ticket price*)
- In general substitutability depends on the “semantic intersection” of the senses of the two lexemes and, in some cases, also by social factors (*father/dad*)

# Hyponymy / Hypernymy

- **Hyponymy** is a relationship between two lexemes (more precisely two senses) such that one denotes a subclass of the other
  - car, vehicle – shark, fish – apple, fruit
  - The relationship is not symmetric
    - The more specialized concept is the hyponym of the more general one
    - The more general concept is the hypernym of the more specialized one
  - Hyponymy (hypernymy) is the basis for the definition of a taxonomy ( a tree structure that defines inclusion relationships in an object ontology) even if it is not properly a taxonomy
    - The definition of a formal taxonomy would require a more uniform/rigorous formalism in the interpretation of the inclusion relationship
    - However the relationship defines an inheritance mechanism of the properties from the ancestors of a given concept in the hierarchy

# Wordnet

- It is a lexical database for English (versions for other languages are available) organized as a **semantic network of senses**
  - It represents nouns, verbs, adjectives, and adverbs but it does not include functional terms in the closed classes (prepositions, conjunctions, etc.)
  - The lexemes are grouped into sets of cognitive synonyms (synset), each representing a distinct concept
  - A set of senses (synset) is associated to each lexeme (unique orthographic form)
  - Synsets are linked by conceptual/semantic and lexical relationships
  - Wordnet consists in lexicographic files, an application to load these files into a database and a library of search and browsing functions to visualize and access the database contents

# Wordnet Statistics

| PoS       | Unique strings | Synset  | pairs word-sense |
|-----------|----------------|---------|------------------|
| Noun      | 117,798        | 82,115  | 146,312          |
| Verb      | 11,529         | 13,767  | 25,047           |
| Adjective | 21,479         | 18,156  | 30,002           |
| Adverb    | 4,481          | 3,621   | 5,580            |
| Total     | 155,287        | 117,659 | 206,941          |

- Nouns have an average of 1.24 senses, verbs 2.17, adjectives 1.40, adverbs 1.25
- The actual total number of distinct strings is 147,278 (the same string can belong to more than one PoS class)

# Synset

- A synset is a set of synonyms that define a concept or word meaning
  - About half of the synsets (~54%) contains only one term, about one third (~29%) 2 terms, about 10% 3 terms
  - An annotation (gloss) explaining the meaning is associated to each synset (especially to those containing a single term)
    - A synset contains ~1.75 terms in average)

2 senses of teacher

Sense 1

synset **teacher#1, instructor#1** -- (a person whose occupation is teaching)  
=> educator#1, pedagogue#1, pedagog#1 -- (someone who educates young people)

Sense 2

synset **teacher#2** -- (a personified abstraction that teaches; "books were his teachers"; "experience is a demanding teacher")  
=> abstraction#1, abstract#1 -- (a concept or idea not associated with any specific instance; "he loved her only in the abstract--not in person")

# Synset – verb example

5 senses of **derive**

Sense 1

**deduce#1, infer#1, deduct#3, derive#1** -- (reason by deduction; establish by deduction)  
=> reason#1, reason out#1, conclude#1 -- (decide by reasoning; draw or come to a conclusion; "We reasoned that it was cheaper to rent than to buy a house")

Sense 2

**derive#2, gain#1** -- (obtain; "derive pleasure from one's garden")  
=> obtain#1 -- (come into possession of; "How did you obtain the visa?")

Sense 3

**derive#3** -- (come from; "The present name derives from an older form")  
=> evolve#2 -- (undergo development or evolution; "Modern man evolved a long time ago")

Sense 4

**derive#4, educe#2** -- (develop or evolve from a latent or potential state)  
=> make#3, create#1 -- (make or cause to be or to become; "make a mess in one's office"; "create a furor")

Sense 5

**derive#5, come#18, descend#2** -- (come from; be connected by a relationship of blood, for example; "She was descended from an old Italian noble family"; "he comes from humble origins")

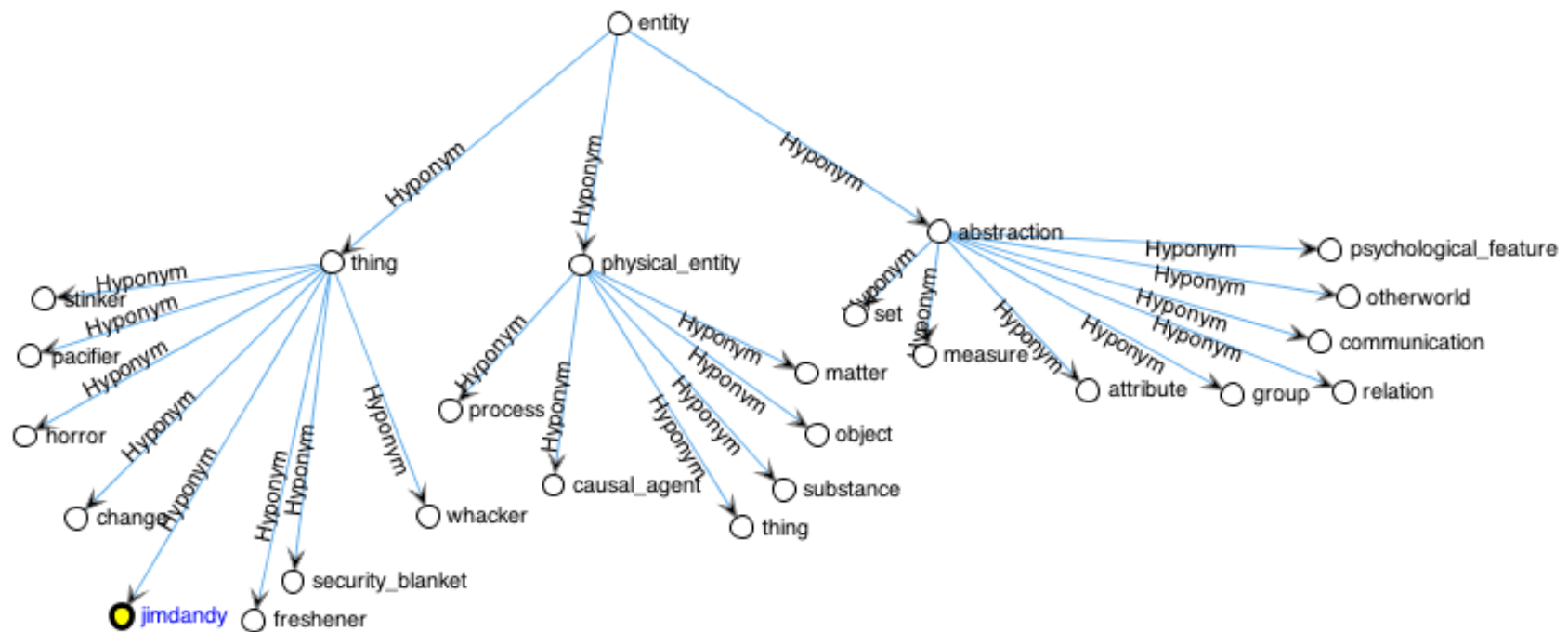
# Names

- Names are organized in a hierarchy of specializations (hyponyms) and generalizations (hypernyms)
  - In the 3.0 version there is a unique root category {entity} referred to as **unique beginner** whereas in the previous versions there are mode unique beginners (25 in version 1.7.1)

**vase#1** -- (an open jar of glass or porcelain used as an ornament or to hold flowers)  
 => **jar#1** -- (a vessel (usually cylindrical) with a wide mouth and without handles)  
 => **vessel#3** -- (an object used as a container (especially for liquids))  
 => **container#1** -- (any object that can be used to hold things (especially a large metal boxlike object of standardized dimensions that can be loaded from one form of transport to another))  
 => **instrumentality#3, instrumentation#1** -- (an artifact (or system of artifacts) that is instrumental in accomplishing some end)  
 => **artifact#1, artefact#1** -- (a man-made object taken as a whole)  
 => **whole#2, unit#6** -- (an assemblage of parts that is regarded as a single entity; "how big is that part compared to the whole?"; "the team is a unit")  
 => **object#1, physical object#1** -- (a tangible and visible entity; an entity that can cast a shadow; "it was full of rackets, balls and other objects")  
 => **physical entity#1** -- (an entity that has physical existence)  
 => **entity#1** -- (that which is perceived or known or inferred to have its own distinct existence (living or nonliving))

# Name hierarchy

- The first three levels of the name hierarchy starting from the unique beginner {entity} (root) in Wordnet 3.0



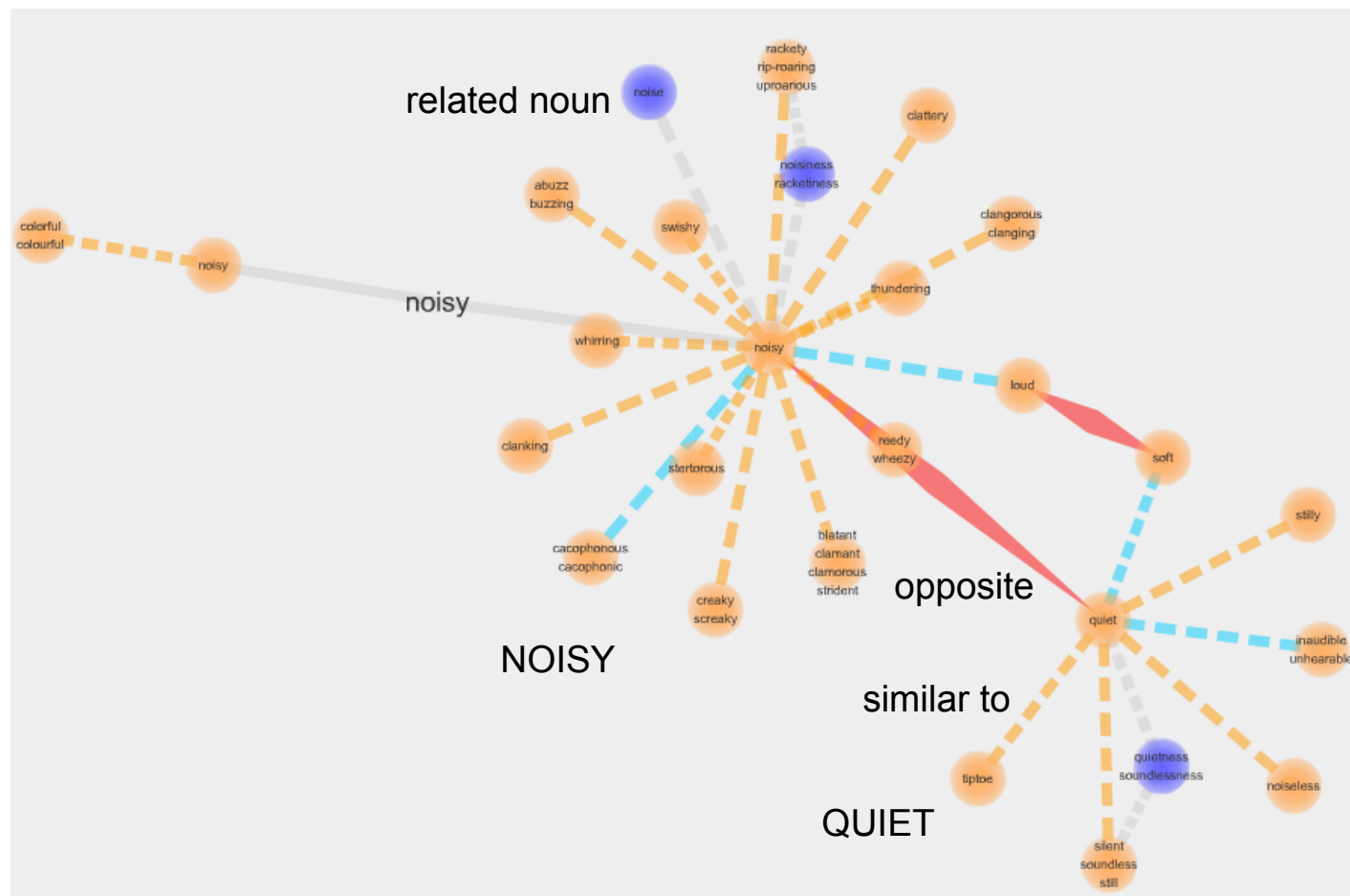
# Wordnet relationships- names & verbs

- For nouns the following relationships are provided among synsets
  - **Hyperonymy** - X is a kind of Y (car → vehicle)
  - **Hyponymy** – Y is a kind of X (vehicle → car)
  - **Coordinate terms**– Y is a coordinate term of X is X and Y share a common hyperonym (car and motorcycle)
  - **Holonymy** – X is part of Y (wheel → bicycle)
  - **Meronymy** – Y is part of X (bicycle → wheel)
- For verbs the following relationships are provided among synsets
  - **Hyperonymy** – the activity of X is a kind of Y (to see → to perceive)
  - **Troponymy** – the activity Y executes X in some sense (to eat → to devour)
  - **Entailment** – Y is required to perform X (to snore→to sleep)
  - **Coordinate terms**– Terms share a common hyperonym (to hear-to see as cases of to perceive)

# Wordnet relations- adjectives & adverbs

- Words can be linked to other words by lexical relationships such as **antonymy** (words that have opposite meanings)
  - good  $\Leftrightarrow$  bad, day  $\Leftrightarrow$  night, exit  $\Leftrightarrow$  entrance
- For adjectives the following relationships are defined
  - **Related nouns**– (noisy  $\rightarrow$  noise)
  - **Similar to**– (noisy  $\rightarrow$  clanking)
    - The **descriptive adjectives** are organized into groups containing a main synset (head) and satellite synsets. Each group is organized around a pair (sometimes a triple) of antonyms corresponding to the main terms. The satellite synsets are those linked by the “*Similar to*” relationship.
    - Relational adjectives are used to categorize the noun and they have neither a group structure nor an antonym (e.g. musical, nervous)
- For the adverbs the following relationships are defined
  - **Base adjective**– (slowly  $\rightarrow$  slow)

## Example – descriptive adjective



# Word sense disambiguation

- Word sense disambiguation (WSD) is the task of selecting the correct sense for a word in a given sentence
  - This problem has to be faced for words having more meanings
  - It requires a dictionary listing all the possible senses for each word
  - It can be faced for each single word or jointly for all the words in the sentence (all the meaning combinations should be considered)

I ate a cold **dish**  
I washed a dirty **dish**  
The *served* a cold **dish**

- Several approaches to WSD have been proposed
  - (Machine Readable) Dictionary and knowledge-based, Machine Learning Supervised methods, Semi-supervised and Unsupervised methods

# Supervised learning

- WSD can be approached as a classification task
  - The correct sense is the class to be predicted
  - The word is represented by a set (vector) of features to be processed as the classifier input
    - Usually the feature include a representation of the word to be disambiguated (target) and of its context (a given number of words at the left and the right of the target word)
    - The word itself, the word stem, the word PoS can be exploited as features
  - The classifier can be learnt from examples given a labeled dataset
  - Different models can be exploited to implement the classifier (Naïve Bayes, neural networks, decision trees...)
  - The limitation of the learning based approach is scalability when a large number of labeled examples is required

# Naïve Bayes

- The bayesian approach aims at maximizing of the probability of sense  $s$  given the feature vector  $f_w$  describing the target word

$$\hat{s} = \operatorname{argmax}_{s \in S} p(s|f_w) = \operatorname{argmax}_{s \in S} \frac{p(f_w|s)p(s)}{p(f_w)}$$

- With the simplifying assumption that the feature vector entries (words in context) are independent of each other  $p(f_w|s)$  can be written as

$$p(f_w|s) = \prod_{j=1}^n p(f_j|s)$$

- the probabilities  $p(f_j|s)$  model the statistics for distribution of feature  $j$  (e.g. a given word) in the context of word  $w$  when having the sense  $s$
- $p(s)$  is the a priori probability of each sense of the word

# Dictionary-based methods

- A dictionary can provide useful information about the contexts related to the word senses (glosses)
  - A simple approach is the Lesk algorithm (1986)
    - The algorithm computes the intersection among the glosses associated to the different meanings of the words in the sentence
    - The combination yielding the maximum overall intersection is selected (the complexity is combinatorial in the number of senses)

pine cone

PINE

1. kinds 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

$$\text{pine}_1 \cap \text{cone}_1 = 0$$

$$\text{pine}_1 \cap \text{cone}_2 = 0$$

$$\text{pine}_1 \cap \text{cone}_3 = 2$$

$$\text{pine}_2 \cap \text{cone}_1 = 0$$

$$\text{pine}_2 \cap \text{cone}_2 = 0$$

$$\text{pine}_2 \cap \text{cone}_3 = 0$$

# Limitations of the Lesk algorithm

- The Lesk algorithm yields a 50-70% accuracy
  - The main limitation is its dependence on the quality of the glosses/ examples provided for each sense in the dictionary since they are usually short and do not carry enough information to train a classifier
    - The words in the context and their definition should share a significant intersection (they should share the maximum number of terms)
  - The coverage can be improved by adding the words related to the target but not already contained in the glosses
    - for example the words of the definitions containing the target word only when the actual sense of the target word is clear in that context
  - In the computation of the intersection/similarity among the context more flexible measure can be exploited
    - Correlation with TermFrequency-InverseDocumentFrequency weights in order to reduce the importance of most common words

# Word similarity

- Synonymy is a kind of strict similarity
  - It defines a complete equivalence (substitutability) in given contexts
- A word similarity can be defined a kind of “semantic distance”
  - Definition on a given Thesaurus (e.g. Wordnet)
    - Words (senses) are linked by different relationships in a thesaurus
    - A semantic distance can be defined by the length of the minimum path leading from the first word to the other using the links of a give relationship

$$\text{sim}(w_1, w_2) = -\log \text{pathlength}(s(w_1), s(w_2))$$

- Definition on the statistical distribution
  - Probability of finding the two words in similar contexts
  - Words are represented by N-dimensional vector and the similarity/distance is computed in this space