

1.10 - Semantics



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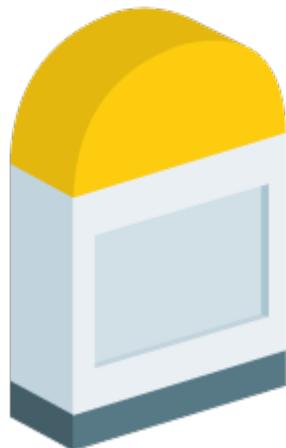


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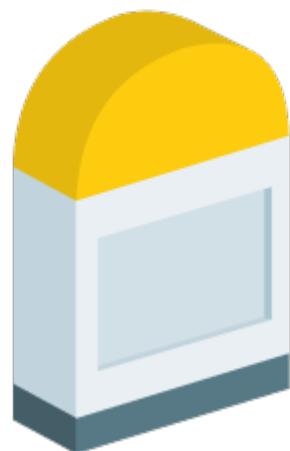
- Semiotic Triangle, From Syntax to Semantics
- Syntax-driven semantic analysis: semantic attachments
- First-Order Logic, lambda calculus for NLP
- Lexical Semantics, lemmatization, word senses, monosemy vs. polysemy, homonymy, metonymy, metaphor, personification, synesthesia, enumerative approach, generative approach.
- WordNet, BabelNet
- Word Sense Disambiguation, Entity Linking, Ontology Learning
- Q&A

Milestones in NLP

today topic is fundamental for the three areas.



rule-based systems

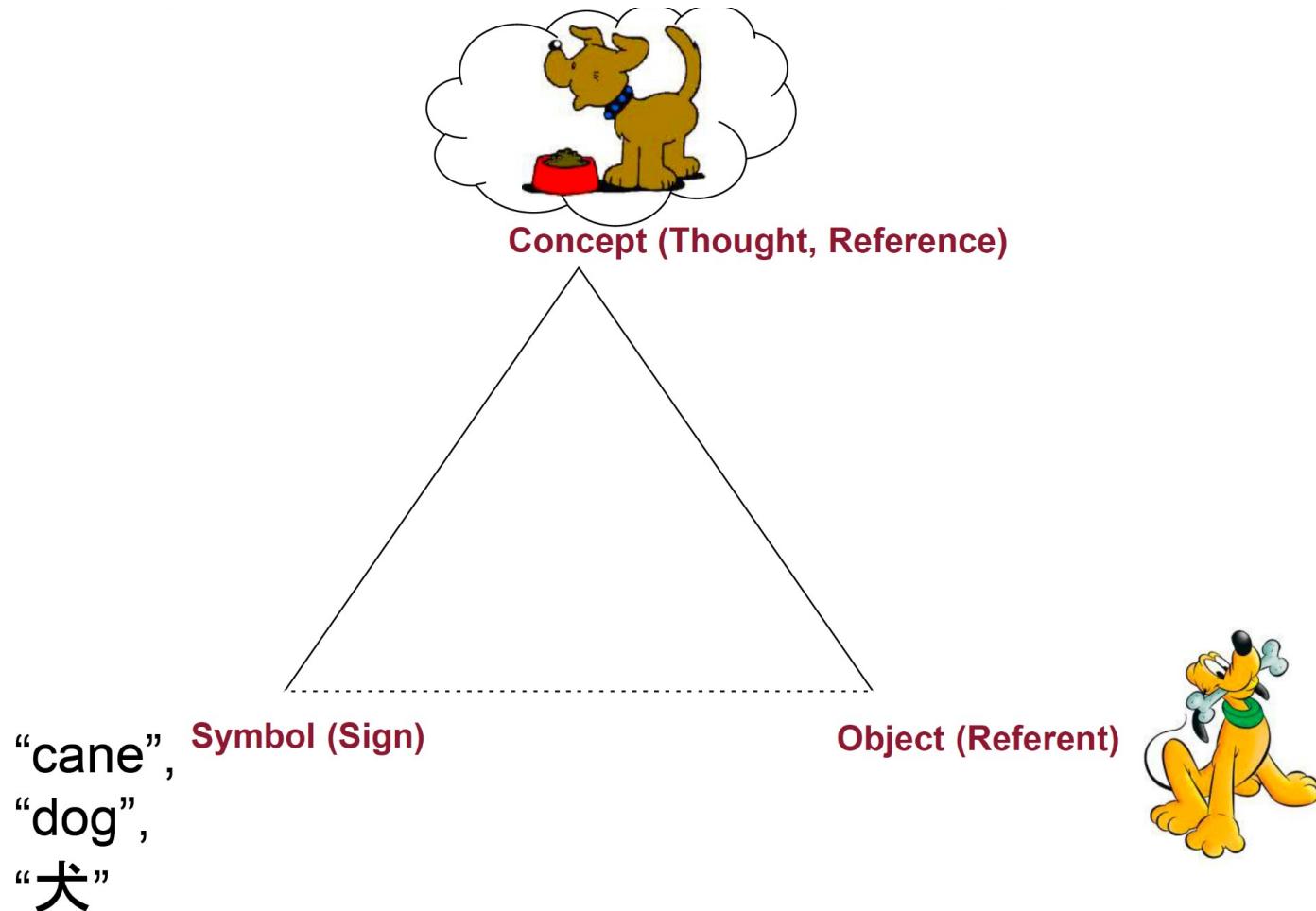


**statistical classical machine learning
models**



deep learning models

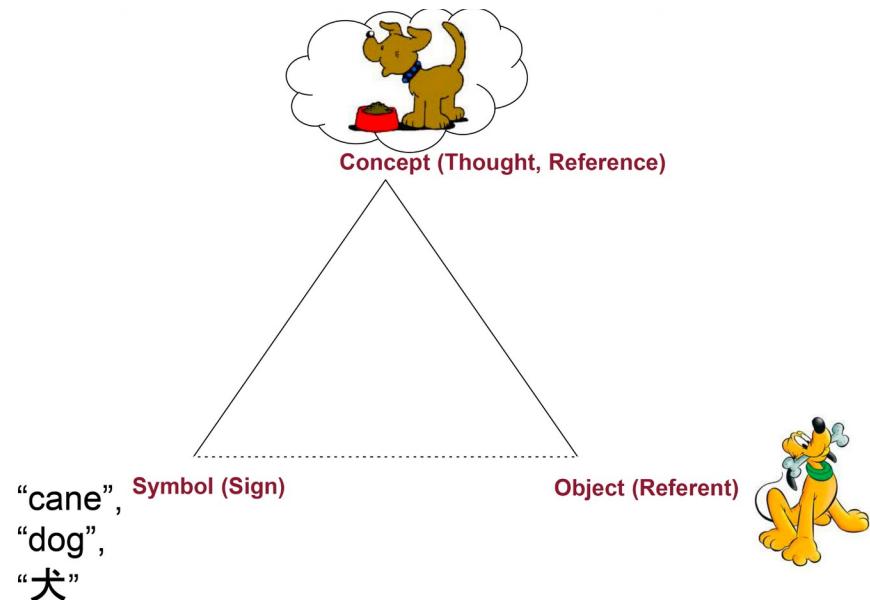
Semantics: Semiotic Triangle (Triangle of meaning)



[Ogden&Richards, 1923]

Semantics: Semiotic Triangle (Triangle of meaning)

The **triangle of meaning** is a model of communication that indicates the **relationship** among a **thought**, **symbol**, and **referent**, and highlights the indirect relationship between the symbol and the referent. The model explains how for any given symbol there can be many different referents, which can lead to misunderstanding.



[Ogden&Richards, 1923]

From Syntax to Semantics

We are now moving from syntax to semantics:

Book the dinner flight

From Syntax to Semantics

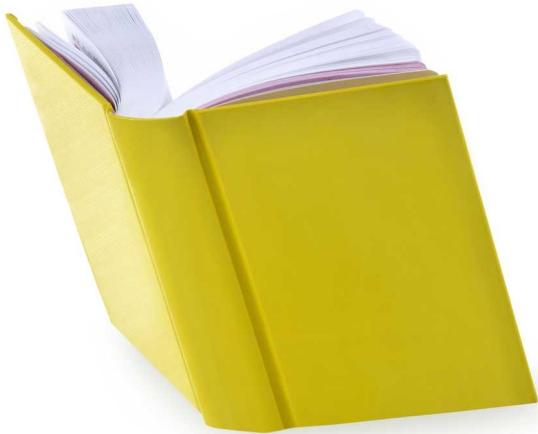
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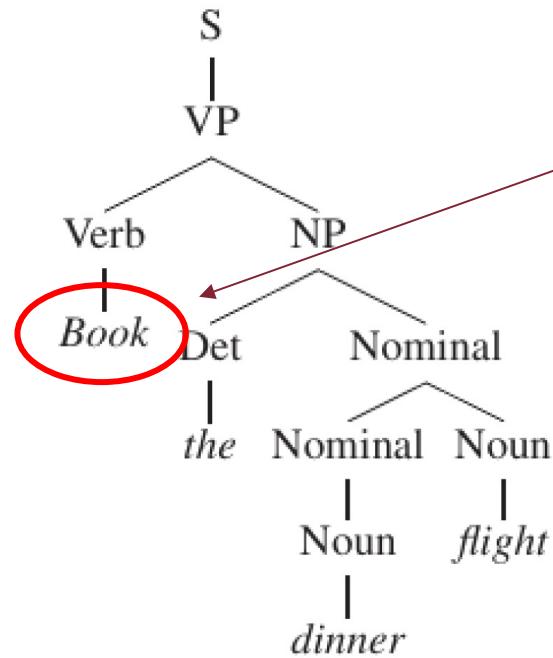


What is the correct semiotic triangle for the word book in this context? -> What is the semantics (the word meaning) for the word book in this context.

From Syntax to Semantics

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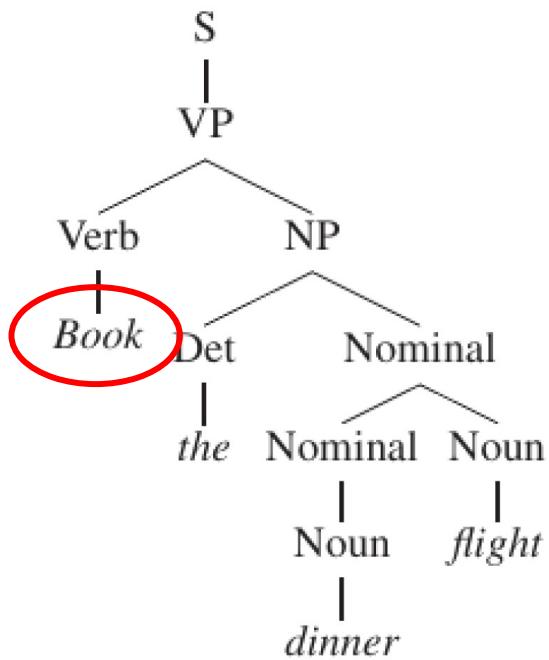


we can understand that,
in this context, book is a
Verb

From Syntax to Semantics

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Book the dinner flight



if we have a sense
inventory

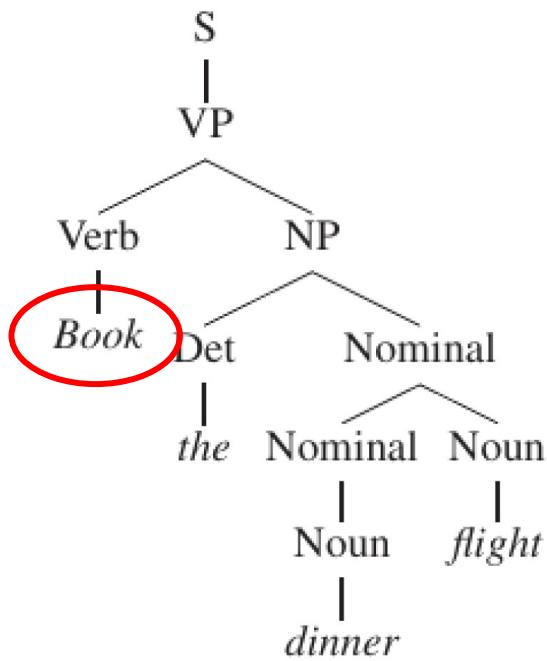
Verb

- S: (v) book (engage for a performance) "Her agent had booked her for several concerts in Tokyo"
- S: (v) reserve, hold, book (arrange for and reserve (something for someone else) in advance) "reserve me a seat on a flight"; "The agent booked tickets to the show for the whole family"; "please hold a table at Maxim's"
- S: (v) book (record a charge in a police register) "The policeman booked her when she tried to solicit a man"
- S: (v) book (register in a hotel booker)

From Syntax to Semantics

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?

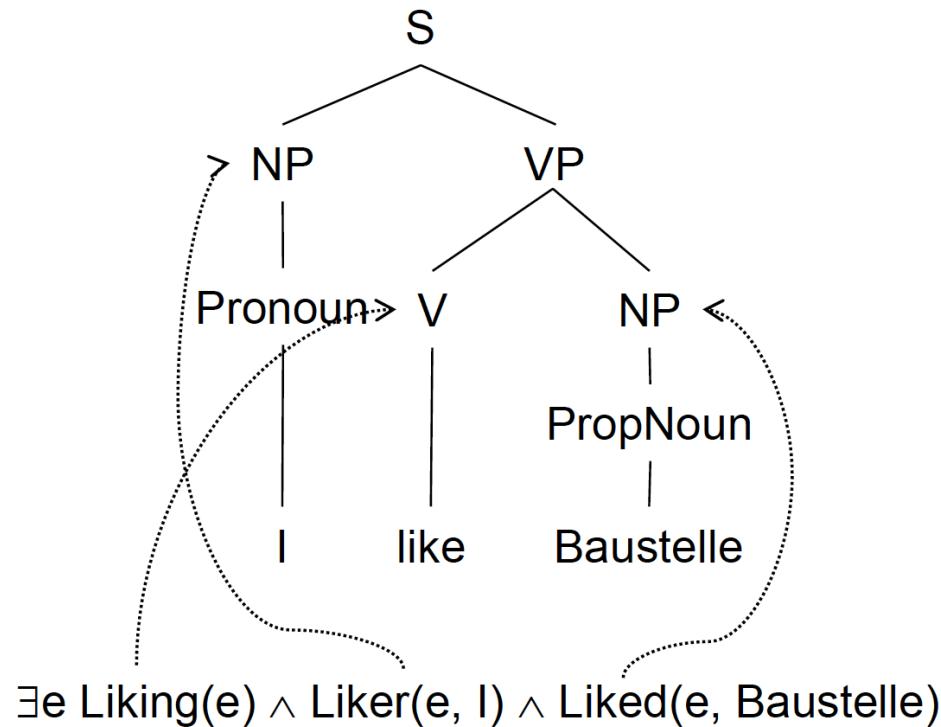
it is a challenging problem to understand the meaning of words in context!

Syntax-driven semantic analysis: Semantic Attachments

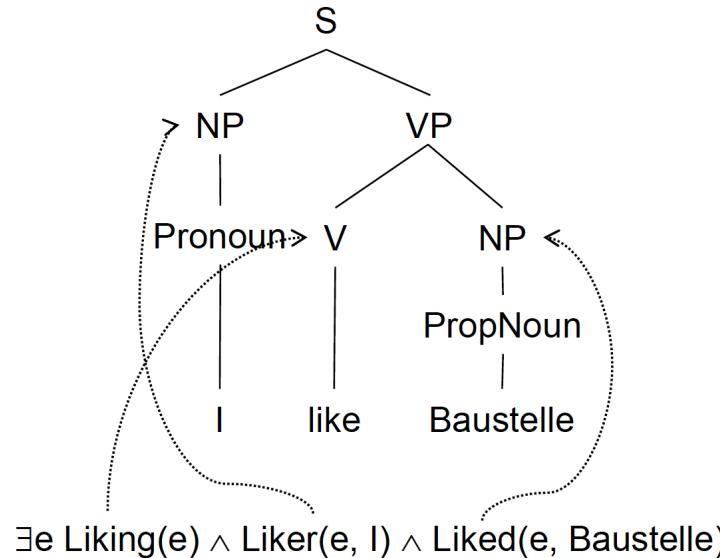
We start from a simplified kind of semantic analysis driven by syntax

- Very complex to use in the real world. Why?

- Example: “I like Baustelle”



Syntax-driven semantic analysis: Semantic Attachments



The assumption here is that we have enough knowledge about this specific example and its parse tree.

But there is an infinite number of such examples and parse trees!

Syntax-driven semantic analysis: Semantic Attachments

We already faced this problem!

E.g., To encode a language, we do not enumerate the strings in the language

- We use a finite formalism that can be used to generate all the strings in the language
- We can do the same for semantics and attach meaning to productions
 - Rule-to-rule hypothesis

Syntax-driven semantic analysis: Semantic Attachments

We can augment CFGs with semantic attachments:

$$A \rightarrow \beta_1 \dots \beta_n \{ f(\beta_1.\text{sem}, \dots, \beta_n.\text{sem}) \}$$

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What can we attach to our context-free rules?

- Our function **f** could be a program fragment
 - The semantic interpretation would be obtained as a result of the bottom-up execution of such fragments
 - Semantic results would be stored for each non-terminal
 - Remember Lex and Yacc?
- But:
 - We cannot (immediately) use logical expressions
 - How can we design grammars with such programs?

Syntax-driven semantic analysis: Semantic Attachments

One option is to adopt **First-Order Logic (FOL)**,

- you all know what it is...but just to be sure...

```
Formula → AtomicFormula
          | Formula Connective Formula
          | Quantifier Variable, ... Formula
          | ¬ Formula
          | (Formula)

AtomicFormula → Predicate(Term, ...)

Term → Function(Term, ...)
       | Constant
       | Variable

Connective → ∧ | ∨ | ⇒

Quantifier → ∀ | ∃

Constant → A | VegetarianFood | Maharani...
Variable → x | y | ...
Predicate → Serves | Near | ...
Function → LocationOf | CuisineOf | ...
```

Syntax-driven semantic analysis: Semantic Attachments

With such a **First-Order Logic (FOL)** specification we can express facts in the following form:

Formula → *AtomicFormula*

- | *Formula Connective Formula*
- | *Quantifier Variable, ... Formula*
- | \neg *Formula*
- | $($ *Formula* $)$

AtomicFormula → *Predicate(Term, ...)*

Term → *Function(Term, ...)*

- | *Constant*
- | *Variable*

Connective → \wedge | \vee | \Rightarrow

Quantifier → \forall | \exists

Constant → *A* | *VegetarianFood* | *Maharani* ...

Variable → *x* | *y* | ...

Predicate → *Serves* | *Near* | ...

Function → *LocationOf* | *CuisineOf* | ...

Le Scalette's location is Nemi

$\text{LocationOf}(\text{LeScalette}, \text{Nemi})$

Le Scalette is a restaurant and Le Scalette's location is Nemi

$\text{Restaurant}(\text{LeScalette}) \wedge \text{LocationOf}(\text{LeScalette}, \text{Nemi})$

There exists a restaurant in Nemi?

$\exists x \text{ Restaurant}(x) \wedge \text{LocationOf}(x, \text{Nemi})$

All pizzerias serve pizzas

$\forall x \text{ Pizzeria}(x) \Rightarrow \text{Serves}(x, \text{Pizza})$

Syntax-driven semantic analysis: Semantic Attachments

In order to abstract away from fully specified FOL, we can use the **lambda notation**

It extends the syntax of FOL to include expressions of the form:

$$\lambda x.P(x)$$

You can apply these expressions to other logical terms to yield new FOL expressions

- For instance: $(\lambda x.\text{Restaurant}(x))(\text{LeScalette}) \rightarrow \text{Restaurant}(\text{LeScalette})$



This operation is called **beta reduction**

Lambda Calculus



Alonzo Church, 1930s

- Lambda Calculus consists of lambda terms:
 - **Variables:** A variable x is a **lambda term**
 - **Abstraction:** If t is a lambda term and x a variable, $\lambda x.t$ is a lambda term:
 - λx represents an anonymous function with a given input x
 - λ binds x in t
 - **Application:** if t and s are lambda terms, then ts is a lambda term
- ...and extra operations:
 - **Alpha conversion:** bound variables can be renamed
 - **Beta reduction:** an application of the form $(\lambda x.t)s$ reduces to the term $t[x := s]$

Lambda Calculus: examples

- Identity: $\lambda x.x$
- Constant: $\lambda x.y$
- Square function: $\lambda x.x^*x$
 - What is the square of 4? $(\lambda x.x^*x) 4 = 4^*4$
- Multiple-argument functions: $\lambda x.\lambda y.x^*y$

Lambda Calculus: Linguistic examples

- We need **constants** to build phrases:
two, red, loves, ate, John, Mary, etc.
- **Connectives and quantifiers:**
or, and, not, if ... then, for all, there exists
- Examples:
 $\text{loves}(\text{John}, \text{Mary})$
 $\exists m. \text{mozzarella}(m) \wedge \text{ate}(\text{John}, m)$
 $\exists m. \text{mozzarella}(m) \wedge \text{blue}(m) \wedge \text{ate}(\text{John}, m)$

Lambda Calculus: modelings events

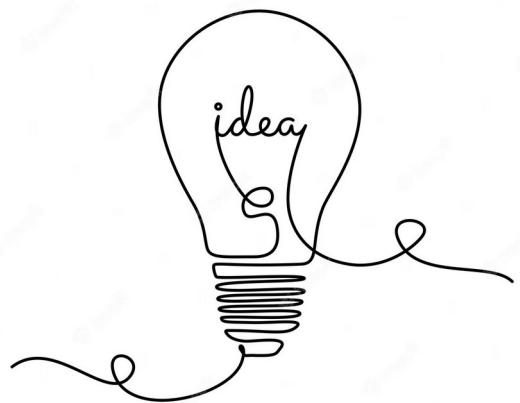
If something happens in a specific moment or period of time, we have an event.

Example:

$$\exists e. \text{past}(e) \wedge \text{act}(e, \text{eating}) \wedge \text{eater}(e, \text{John}) \wedge \exists m \\ (\text{mozzarella}(m) \wedge \text{eatee}(e, m))$$

How to Get Semantics from sentences

1. Parse sentences into parse trees
2. Get the **semantics** for each word
3. Use the parse tree to build the semantics **bottom-up**
(from words to more complex constituents)

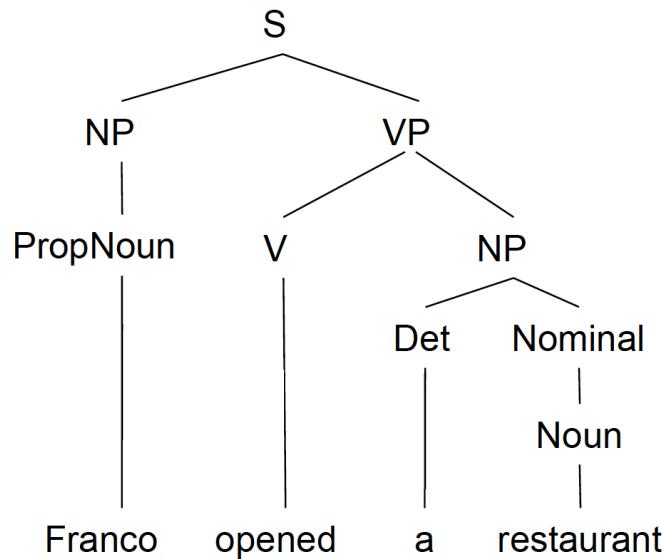


An example excerpt

Grammar Rule	Semantic Attachment
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$Verb \rightarrow closed$	$\{\lambda x. \exists e Closing(e) \wedge Closed(e, x)\}$
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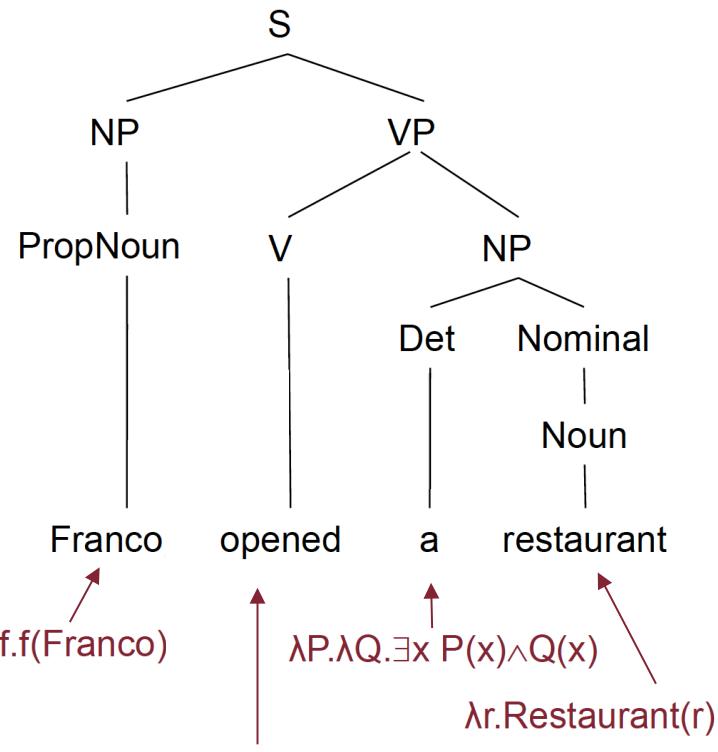
Franco opened a restaurant



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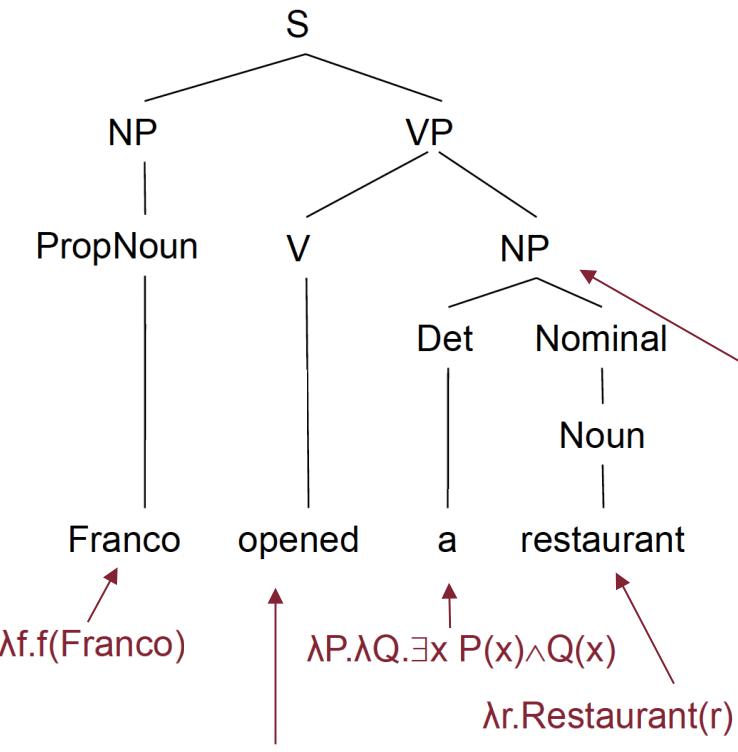
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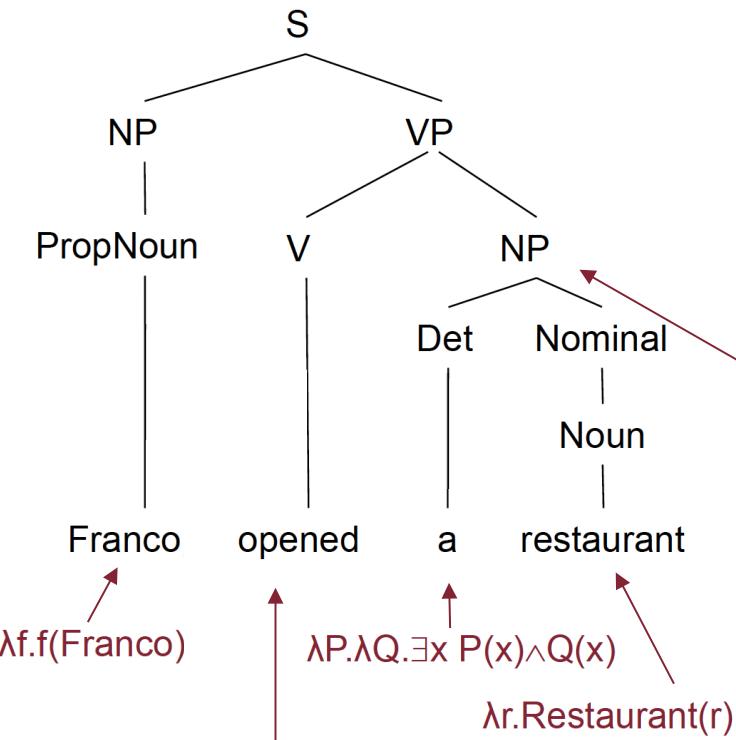
$$\begin{aligned}
 & (\lambda P. \lambda Q. x P(x) Q(x)) \lambda r. Restaurant(r) \\
 & = \lambda Q. x Restaurant(x) Q(x)
 \end{aligned}$$

Beta-reduction

$$\lambda w. \lambda z. w(\lambda x. \exists e Opening(e) \wedge Opener(e, z) \wedge Opened(e, x))$$

An example excerpt

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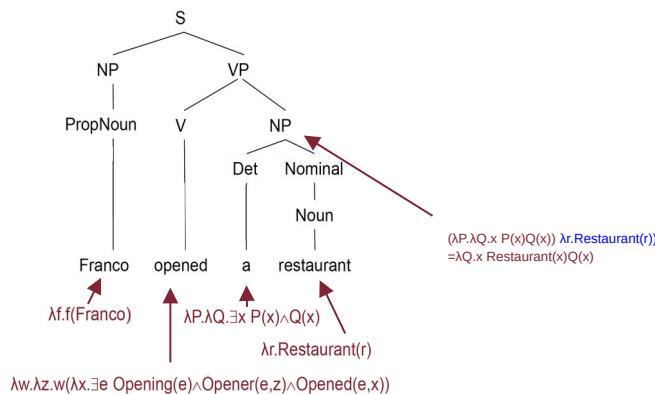
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Beta-reduction

and so on ...

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Sentence:

Franco opened a restaurant

Semantics:

$\exists x \text{ Restaurant}(x) \wedge \exists e \text{ Opening}(e) \wedge \text{Opener}(e, \text{Franco}) \wedge \text{Opened}(e, x)$

Side-effects

1. It's difficult!

Side-effects

1. It's difficult!

2. More dangerous:

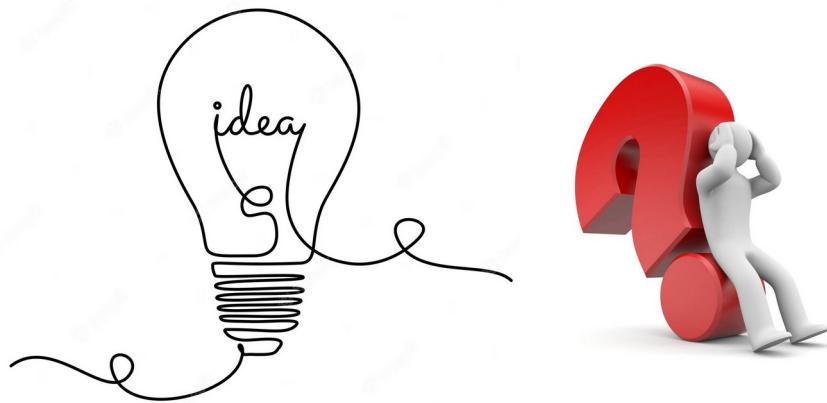
We made a **simplifying assumption** that each terminal (e.g., Opening, Franco, etc.) represents one **meaning**!

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Side-effects

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We made a **simplifying assumption** that each terminal (e.g., Opening, Franco, etc.) represents one **meaning**!

We need to step back to the **meaning of words** before trying to cope with the **meaning of sentences**.

Lexical-semantics

Lexical-semantics

We need to step back to the **meaning of words** before trying to cope with the **meaning of sentences**.

Some definitions:

- **Lexicon:** a finite list of given language words
- **Lemma:** the grammatical base form used in dictionaries
 - (e.g., peach is the lemma of peaches)
- **Word form:** any grammatical form of a word
 - (e.g. eat, eats, ate, etc.)

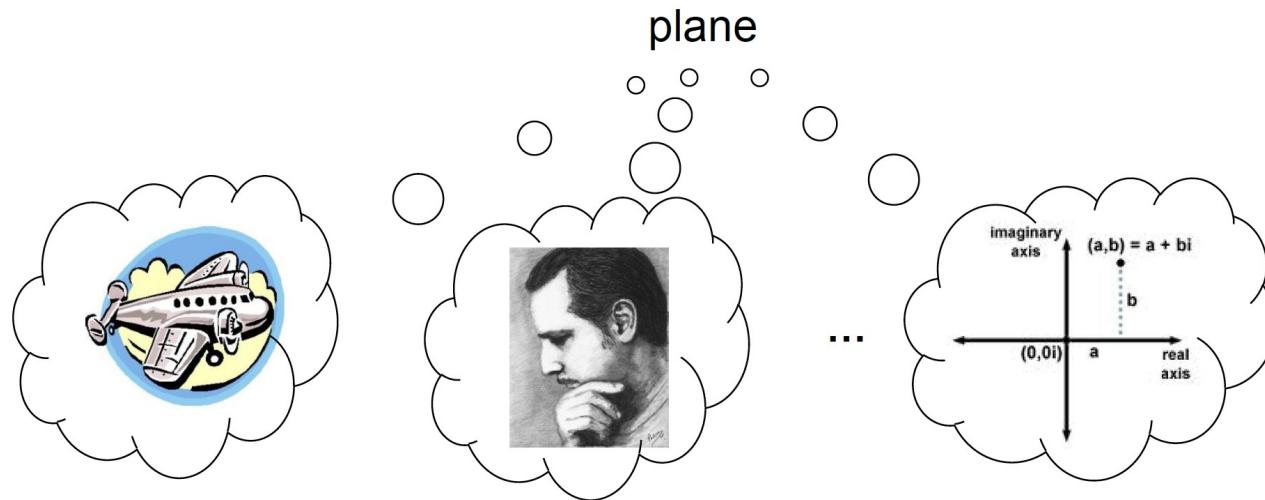
Lexical-semantics: lemmatization

The process of **converting word forms to their lemmas** is called **lemmatization**

- Not always deterministic, it may depend on the context
 - It **bound** the parties to observe neutrality
 - **Lemma:** bind (constrain)
 - The horses were found to **bound** across the meadows
 - **Lemma:** bound (jumping)
- Plus: lemmas are **different** across parts of speech:
 - They closed the **purchase** with a handshake
 - They are about to **purchase** a flat

Word Senses

- The **meaning** of a word depends on the **context** in which it occurs.
- We call each meaning of a word a **word sense**.



Word Senses in context: the “plane” example

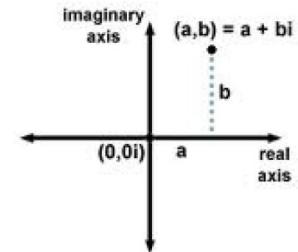
- I am catching the earliest **plane** to Brussels.



- This area probably lies more on the spiritual **plane** than the mental one.



- Let's represent three-dimensional structures on a two-dimensional **plane**



Monosemy vs. Polysemy

- A word is **monosemous** if it has a single sense
 - Internet
 - plant life
 - ...
- **Polysemous** otherwise
 - bank
 - plane
 - bass
 - ...

How do state-of-the-art applications deal with senses?

Google plane

Tutti Immagini Notizie Shopping Video Altro Strumenti Raccolte SafeSearch

disney cartone aereo gioco libro milan antincendio peluche torta

The image shows a Google search results page for the query "plane". The search bar at the top contains the word "plane". Below the search bar, there are tabs for "Tutti", "Immagini" (selected), "Notizie", "Shopping", "Video", "Altro", and "Strumenti". A "Raccolte" (Collections) button and a "SafeSearch" dropdown are also present. Below the tabs, there are several image thumbnails and their corresponding links:

- F Flying Magazine**
How Fast Do Commercial Planes Fly? - FLYING Ma...
- Travel + Leisure**
Here's Why Planes Fly at 36,000 Feet
- Encyclopedia Britannica**
7 Puzzling Plane Disappearances | Britannica
- The New York Times**
What Kind of Plane Am I Flying On? ...
- CN Traveler**
Why You Should Meditate on a Plane (and How to D...
- Business Insider**
Plane With Excess Fuel Flew 'at Low Altitude ...
- M Metro**
Plane dropped 1,200 feet in less than a minute ...
- ThoughtCo**
Plain vs. Plane: How to Choose the Right Word

Homonymy

- That **bank**¹ holds the mortage on my home.
- He sat on the **bank**² of the river and watched the currents.
- The **bank**³ is on the corner of New Oxford Street.
- A huge **bank**⁴ of earth was found close to the entrance.
- **Homonymy**: two senses share an orthographic form (e.g., bank), but are semantically and ethymologically unrelated (**different lemmas!**)



Metonymy

- Metonymy: referring to one aspect of a thing instead of the thing itself or vice versa
 - Have another glass



Metaphor

- **Metaphor:** referring to something which is **similar** to what is meant
 - An **ocean** of trouble



Personification

- Personification (pathetic fallacy): assigning **human features** to things or animals
 - The **crying** fountain



Synesthesia

- **Synesthesia**: referring to a sensation in a sense modality different from the normal expected one
 - The **black yell**



Sense Distinctions

The problem is about the word sense representation:

- How to represent word senses?
 - Can we **enumerate** the senses of a word?



- Can we **generate** the senses of a word?
 - E.g. kiwi is a fruit, the purpose of kiwi is to be eaten, flesh is part of a kiwi, etc.

“**Kiwi** is my mother tongue, but I also speak all other English languages”

Word Senses: the enumerative approach

A fixed **sense inventory** enumerates the range of all possible meanings of a word.



Word Senses: the enumerative approach

A fixed **sense inventory** enumerates the range of all possible meanings of a word.

Example: *knife*

1. a cutting tool composed of a blade with a sharp point and a handle
2. an instrument with a handle and blade with a sharp point used as a weapon



Word Senses: the enumerative approach

A fixed **sense inventory** enumerates the range of all possible meanings of a word.

Example: *knife*

1. a cutting tool composed of a blade with a sharp point and a handle
2. an instrument with a handle and blade with a sharp point used as a weapon

Words in context are assumed to **select/activate** one of these senses in each context



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Words in context are assumed to **select/activate** one of these senses in each context

- She chopped the vegetables with a chef's knife
➡ KNIFE as a **CUTTING TOOL**
- A man was beaten and cut with a knife.
➡ KNIFE as a **WEAPON**

Word Senses: the enumerative approach

A fixed **sense inventory** enumerates the range of all possible meanings of a word.

Example: *knife*

1. a cutting tool composed of a blade with a sharp point and a handle
2. an instrument with a handle and blade with a sharp point used as a weapon

? should we **add a further sense** to the inventory for “a cutting blade forming part of a machine”

? are word senses **application-independent**

Graded word sense assignment [Erk and McCarthy, 2009]

- Similar to **category membership** in the human mind, the meanings of words in context do not always have clear-cut boundaries
- Given a target word, annotators judge the **applicability** of each possible word sense on a **graded scale**

Sentence	Senses							Annotator
	1	2	3	4	5	6	7	
This question provoked arguments in America about the Norton Anthology of Literature by Women, some of the contents of which were said to have had little value as literature.	1	4	4	2	1	1	3	Ann. 1
	4	5	4	2	1	1	4	Ann. 2
	1	4	5	1	1	1	1	Ann. 3

Word Senses: the Generative Approach [Pustejovsky, 1991]

Senses are generated from **rules** which **capture regularities** in the creation of senses

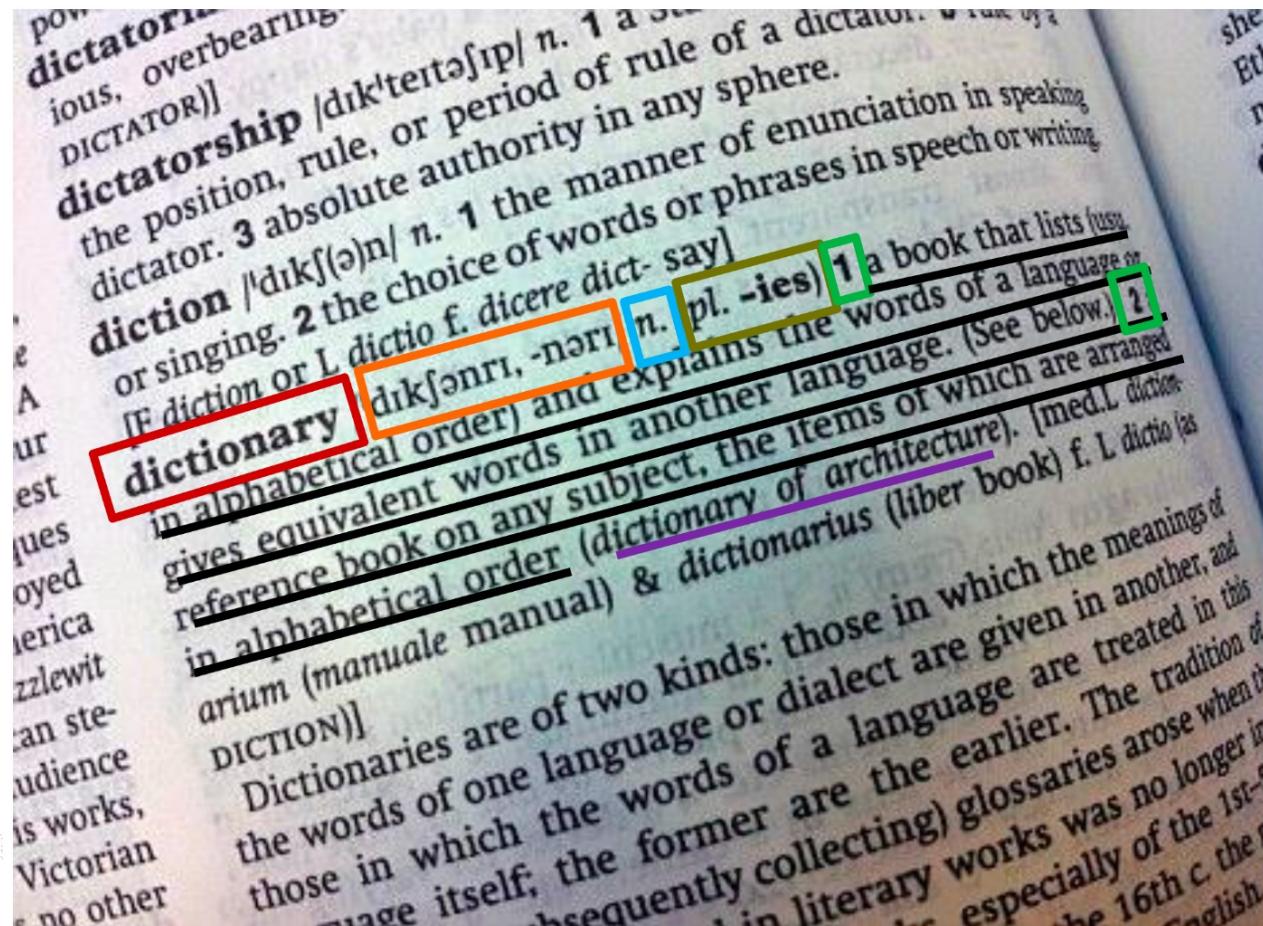
knife	
TYPESTR =	$\left[\text{ARG1} = \boxed{x} \text{ artifact tool} \right]$
ARGSTR =	$\left[\begin{array}{l} \text{D-ARG1} = \boxed{y} \text{ physical object} \\ \text{D-ARG2} = \boxed{w} \text{ human} \\ \text{D-ARG3} = \boxed{z} \text{ human} \\ \text{D-E1} = \boxed{e_1} \text{ transition} \\ \text{D-E2} = \boxed{e_2} \text{ process} \end{array} \right]$
QUALIA =	$\left[\begin{array}{l} \text{FORMAL} = \boxed{x} \\ \text{CONSTITUTIVE} = \{\text{blade, handle, ...}\} \\ \text{TELIC} = \text{cut act}(\boxed{e_2}, \boxed{w}, \boxed{x}, \boxed{y}) \\ \text{AGENTIVE} = \text{make act}(\boxed{e_1}, \boxed{z}, \boxed{x}) \end{array} \right]$

Qualia structures are the **roles** describing the meaning of a lexical element:

- FORMAL: distinguishes a word within a larger domain typing information about the object (e.g., hypernym)
- CONSTITUTIVE: physical properties of a object (e.g., material, parts)
- TELIC: the object's purpose and function
- AGENTIVE: the object's origin or “bringing about”, typically a verb denoting an action which brings the object in existence

How do we encode word senses?

- The **human** way: dictionaries



Lemma
Pronunciation
Part of speech
Morphological info
Senses
Textual definition
or gloss
Usage information

How do we encode word senses?

- The computer way:
 - thesauri: lexical resources that list words grouped together according to similarity of meaning (containing synonyms and sometimes antonyms)

IV. WORDS RELATING TO THE INTELLECTUAL FACULTIES; COMMUNICATION OF IDEAS

III. MEANS OF COMMUNICATING IDEAS

2. Conventional means; written language
Book.

[**Nouns**] booklet; writing, work, volume, tome, opuscule; tract, tractate; livret; brochure, libretto, handbook, codex, manual, pamphlet, enchoridion, circular, publication; chap book.

part, issue, number, livraison; album, portfolio; periodical, serial, magazine, ephemeris, annual, journal.

paper, bill, sheet, broadsheet; leaf, leaflet; fly leaf, page; quire, ream
chapter, section head, article, paragraph, passage, clause.
folio, quarto, octavo; duodecimo, sextodecimo, octodecimo.

How do we encode word senses?

- The computer way:
 - thesauri: lexical resources that list words grouped together according to similarity of meaning (containing synonyms and sometimes antonyms)
 - machine-readable dictionaries (electronic counterparts of paper dictionaries)

The screenshot shows a digital dictionary entry for the word "rhotic, a." in a yellow header bar. The header also includes "ADDITIONS SERIES 15". Below the header, the word is defined as "Phonetics." with sub-links for "Pronunciation", "Etymology", "Quotations", and "Date chart". The definition itself is: "(rɔtɪk) [f. *rhot-*, as in [RHOTACISM](#) *n.*, + [-IC](#).]" A detailed description follows: "Of or pertaining to a variety or dialect of English in which *r* is pronounced not only in pre-vocalic position but also before a consonant or word-finally; characterized by *r*-pronouncing." A historical note at the bottom left discusses the classification of rhoticity based on pronunciation rules, mentioning J.C. Wells, Rhotacism, and various linguistic publications from the 1960s to the 1980s.

How do we encode word senses?

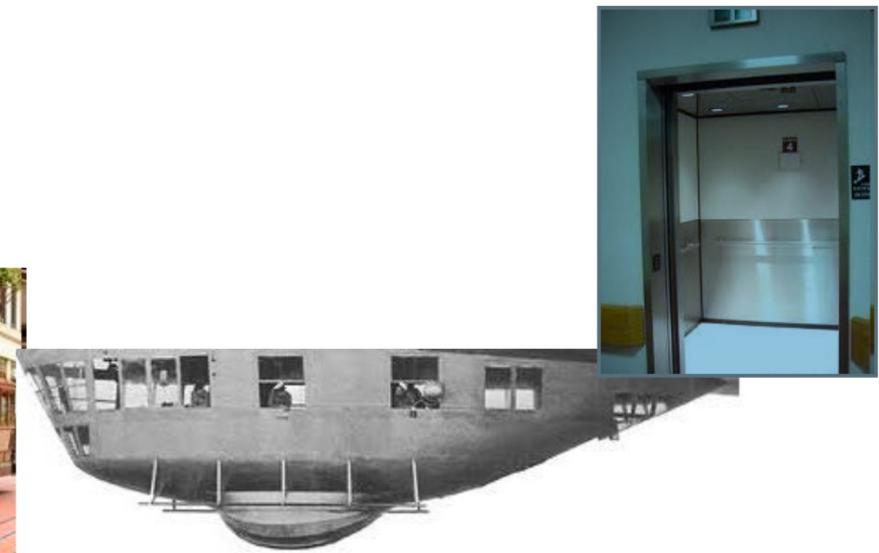
- The computer way:
 - **thesauri**: lexical resources that list words grouped together according to **similarity of meaning** (containing **synonyms** and sometimes **antonyms**)
 - machine-readable dictionaries (electronic counterparts of paper dictionaries)
 - computational lexicons: highly-structured repositories of **syntactic and semantic knowledge** about **words** in a natural language

WordNet [Miller et al. 1990]

- The **most popular computational lexicon** of English
 - Based on psycholinguistic theories
- Concepts expressed as sets of **synonyms** (**synsets**)
 - { car_n¹, auto_n¹, automobile_n¹, machine_n⁴, motorcar_n¹ }
- A **word sense** is a word occurring in a synset
 - **machine_n**⁴ is the fourth sense of noun machine

WordNet: the “car” example

$Senses_{WN}(car_n) = \{ \{car_n^1, auto_n^1, automobile_n^1, machine_n^4, motorcar_n^1\},$
 $\{car_n^2, rail car_n^1, rail way car_n^1, rail road car_n^1\},$
 $\{cable car_n^1, car_n^3\},$
 $\{car_n^4, gondola_n^3\},$
 $\{car_n^5, elevator car_n^1\} \}$.



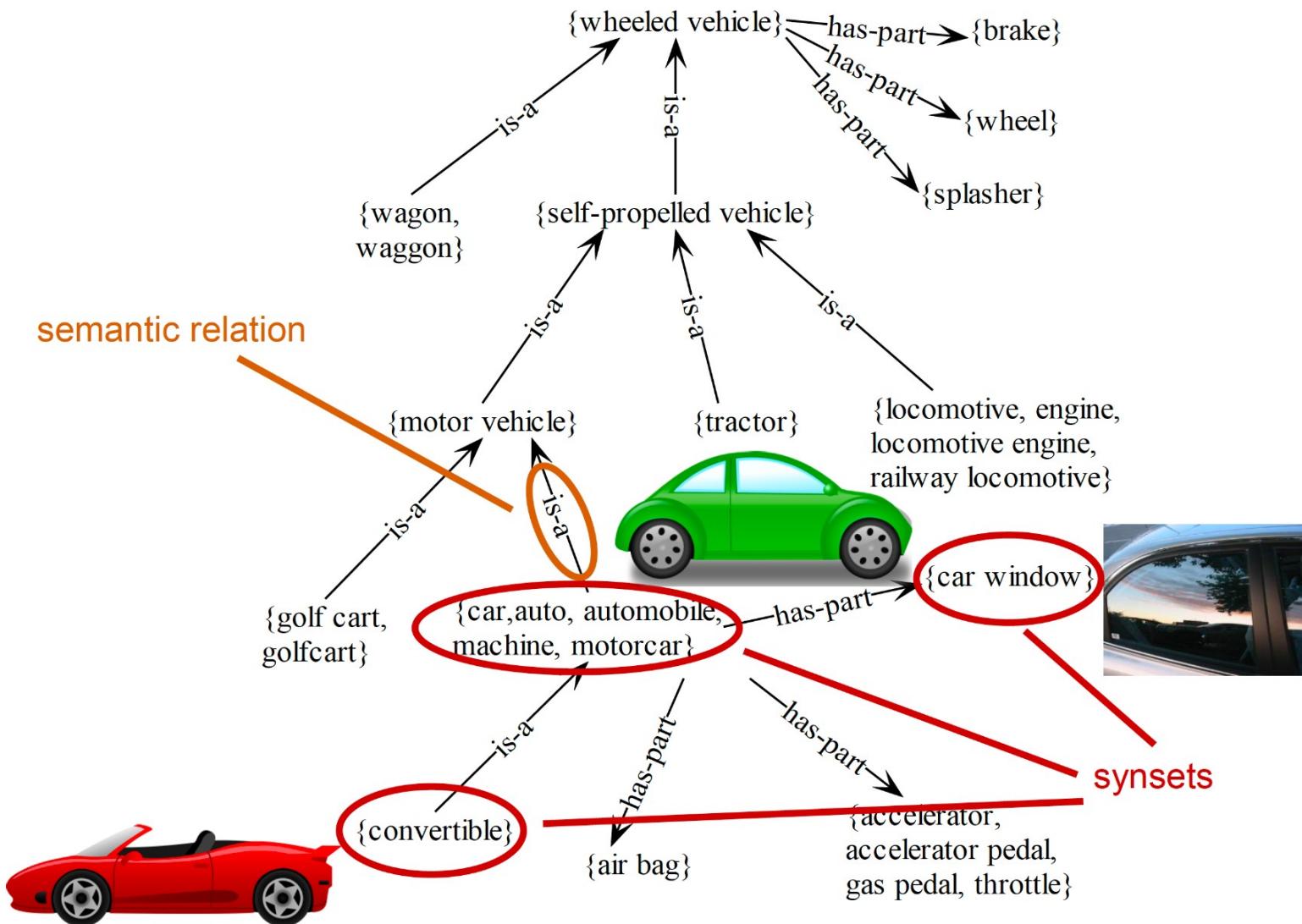
WordNet provides textual definitions

- Called **glosses**
- A textual definition is provided for each synset
- Gloss of car_n^1 : 
 - “a 4-wheeled motor vehicle; usually propelled by an internal combustion engine; ‘he needs a car to get to work’ ”
- Gloss of car_n^2 : 
 - “a wheeled vehicle adapted to the rails of railroad; ‘three cars had jumped the rails’ ”

WordNet encodes relations!

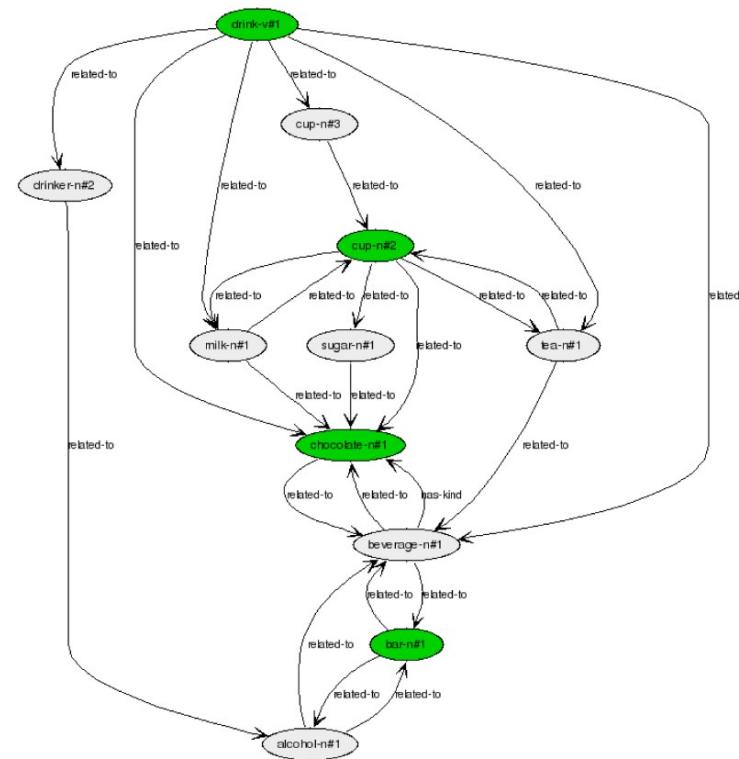
- Semantic relations between synsets
 - **Hypernymy** (car_n^1 is-a motor vehicle $_n^1$)
 - **Meronymy** (car_n^1 has-a car door $_n^1$)
 - **Entailment, similarity, attribute**, etc.
- Lexical relations between word senses
 - **Synonymy** (i.e., words that belong to the same synset)
 - **Antonymy** (good_a^1 antonym of bad_a^1)
 - **Pertainymy** (dental_a^1 pertains to tooth $_n^1$)
 - **Nominalization / derivationally related** (service_n^2 nominalizes serve_v^4)

WordNet as a Graph



But WordNet is more than Simply a Graph!

- It is a semantic network!
- A semantic network is a network which represents semantic relations among concepts
- It is often used as a form of knowledge representation



WordNet [Miller et al. 1990]

<https://wordnet.princeton.edu/>

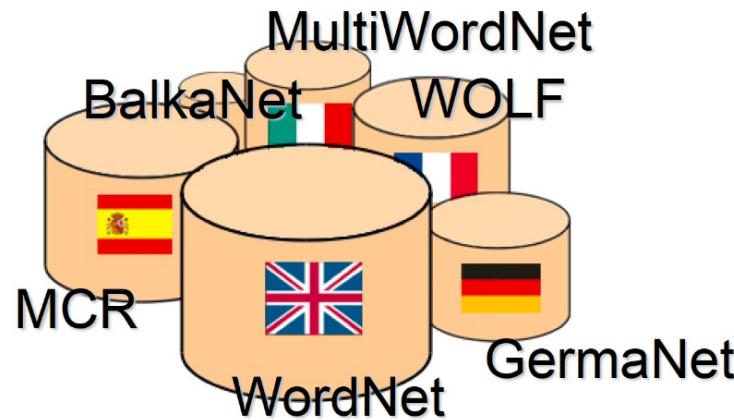
There are several API to programmatically leverage WordNet in your programs.

- Java: (e.g., <https://projects.csail.mit.edu/jwi/>)
- Python: (e.g., <https://www.nltk.org/>)
- ...



Wordnets in other languages

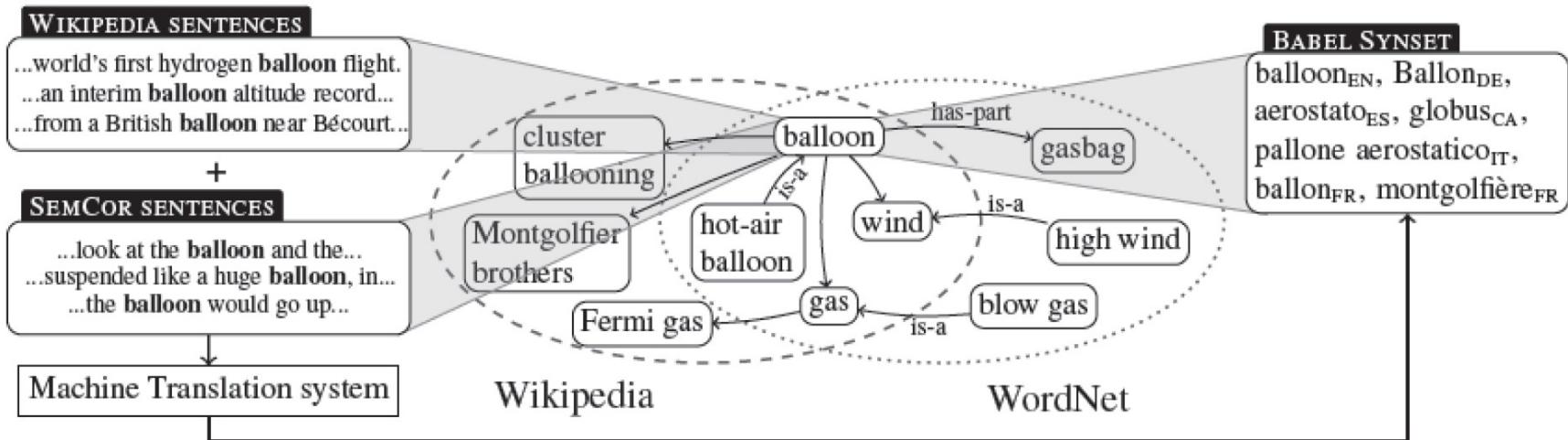
- EuroWordNet [Vossen, 1998]
- BalkaNet [Tufis et al., 2004]
- Multilingual Central Repository [Atserias et al., 2003]



- + Provide **structural knowledge** for languages other than English
- Manually built => limited coverage
- Available only for a small number of languages

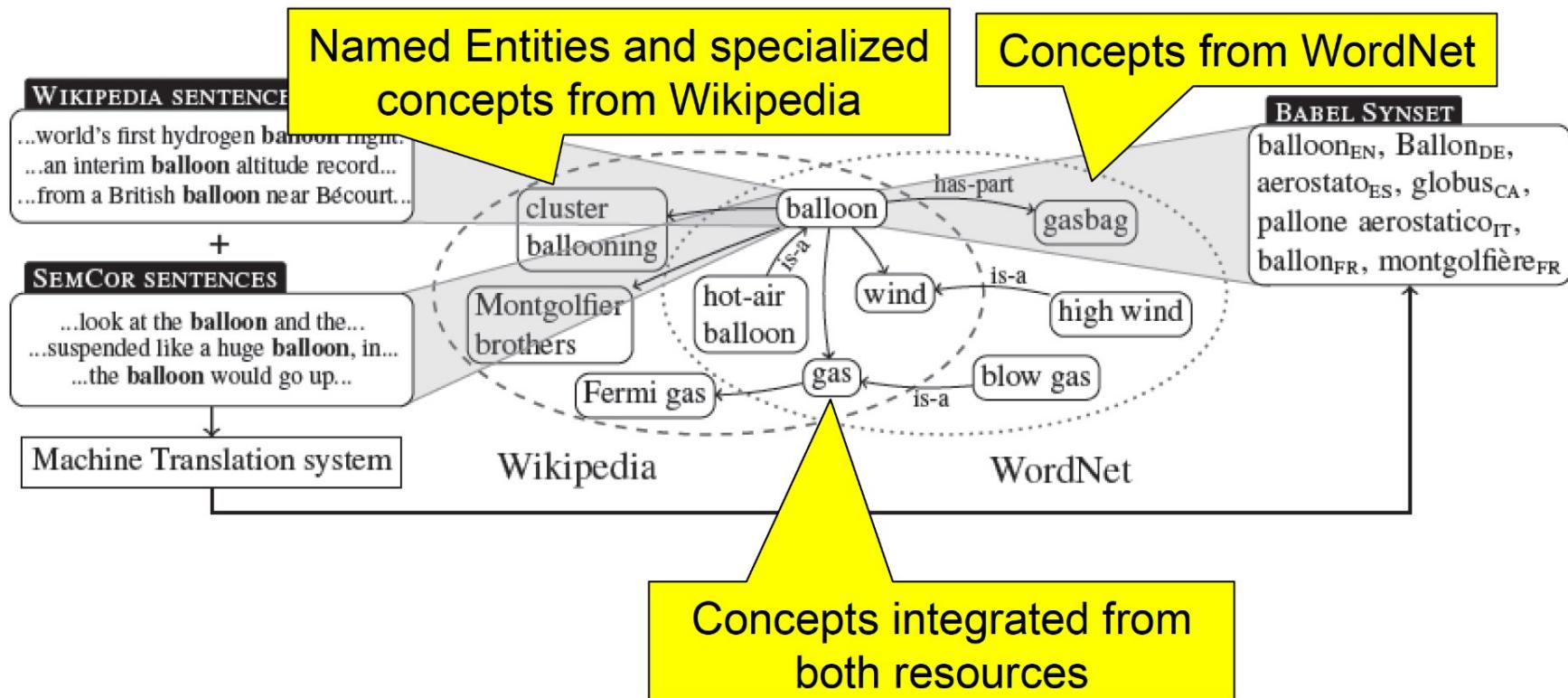
BabelNet [Navigli and Ponzetto, 2010; 2012]

- A wide-coverage multilingual semantic network including both encyclopedic (from Wikipedia) and lexicographic (from WordNet) entries



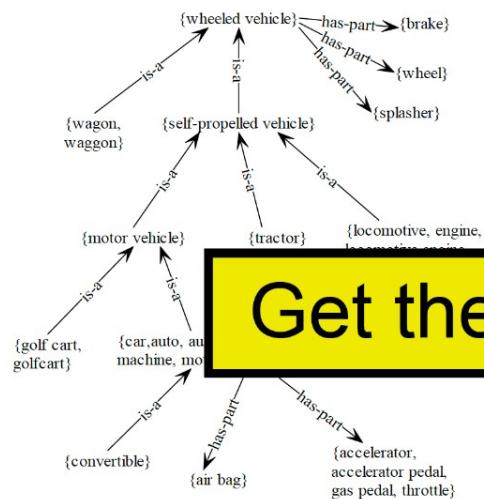
It all started with merging WordNet and Wikipedia [Navigli and Ponzetto, ACL 2010; AIJ 2012]

- A wide-coverage multilingual semantic network including both **encyclopedic** (from Wikipedia) and **lexicographic** (from WordNet) entries

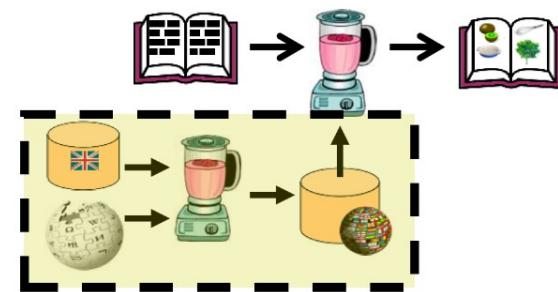


Creating a Multilingual Semantic Network

- Start from two large **complementary** resources:
 - WordNet: full-fledged taxonomy
 - Wikipedia: multilingual and continuously updated



Get the best from both worlds



What is BabelNet?

- A **merger** of resources of different kinds:
 - **WordNet**: the most popular computational lexicon of English
 - **Open Multilingual WordNet**: a collection of open wordnets
 - **WoNeF**: a French WordNet
 - **Wikipedia**: the largest collaborative encyclopedia
 - **Wikidata**: the largest collaborative knowledge base
 - **Wiktionary**: the largest collaborative dictionary
 - **OmegaWiki**: a medium-size collaborative multilingual dictionary
 - **GeoNames**: a worldwide geographical database
 - **FrameNet** lexical units
 - **VerbNet** entries
 - **Microsoft Terminology**: a computer science thesaurus
 - High-quality automatic **sense-based** translations

Anatomy of BabelNet

· bn:00002150n · 명사 ·
개념 · 카테고리: Artificial intelligence,
Cybernetics, Formal sciences, Technology in
society...

EN **artificial intelligence** · AI ·
· artilect · Artifical intelligence ·
Cognitive systems

The branch of **computer science** that deal with writing **computer programs** that can solve problems **creatively**

더 많은 정의

Workers in AI hope to imitate or duplicate intelligence in computers and robots

IS A: computer · intelligence
PART OF: computer

usage examples

카테고리: Intelligence, Cognition, Computer science, Formal sciences, Technology in society... · 사이버네틱스, 과학, 철학, 인공지능...

pronunciation

IT **intelligenza artificiale**

KO **인공지능** · 인공 지능

definition speech

definitions

Con il termine **intelligenza artificiale** si intende generalmente l'abilità di un **computer** di svolgere **funzioni** e ragionamenti tipici della **mente umana**.

더 많은 정의

인공지능은 철학적으로 인간성이나 지성을 갖춘 존재, 혹은 시스템에 의해 만들어진 지능, 즉 인공적인 지능을 뜻한다.

IS A: informatica · intelligenza
PART OF: Scienze cognitive · informatica

PART OF:

IS A: 컴퓨터 과학 · 지능
과학 · 컴퓨터 과학

PART OF: 인지

generalizations and other relations

Why do we need BabelNet?

- **Multilinguality**: the same concept is expressed in tens of languages
- **Coverage**: 284 languages and 16 million entries!
 - **6M** concepts and **7.7M** named entities
 - **119M** word senses
 - **378M** semantic relations (27 relations per concept on avg.)
 - **11M** images associated with concepts
 - **41M** textual definitions
 - **2M** concepts with domains associated



Why do we need BabelNet?

- **Multilinguality**: the same concept is expressed in tens of languages
- **Coverage**: 271 languages and 14 million entries!
- **Concepts and named entities together**: dictionary and encyclopedic knowledge is semantically interconnected
- **"Dictionary of the future"**: semantic network structure with labeled relations, pictures, multilingual synsets
- **Full-fledged taxonomy**: is-a relations are available for both concepts and named entities ([Wikipedia Bitaxonomy](#))
- **Easy access**: Java and HTTP RESTful APIs; SPARQL endpoint (2 billion triples)



Do we need knowledge in some form?



Word Sense Disambiguation (WSD)

Definition:

given a word in context,

(e.g., “I like **planes** and aeronautics in general”)

and a fixed inventory of potential word senses

(e.g., WordNet)

decide the most suitable synsets, if any.

(e.g., plane_n^1)

WordNet Search - 3.1

- [WordNet home page](#) - [Glossary](#) - [Help](#)

Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

Noun

- [S: \(n\) airplane, aeroplane, plane](#) (an aircraft that has a fixed wing and is powered by propellers or jets) "*the flight was delayed due to trouble with the airplane*"
- [S: \(n\) plane, sheet](#) ((mathematics) an unbounded two-dimensional shape) "*we will refer to the plane of the graph as the X-Y plane*"; "*any line joining two points on a plane lies wholly on that plane*"
- [S: \(n\) plane](#) (a level of existence or development) "*he lived on a worldly plane*"
- [S: \(n\) plane, planer, planing machine](#) (a power tool for smoothing or shaping wood)
- [S: \(n\) plane, carpenter's plane, woodworking plane](#) (a carpenter's hand tool with an adjustable blade for smoothing or shaping wood) "*the cabinetmaker used a plane for the finish work*"

Verb

- [S: \(v\) plane, shave](#) (cut or remove with or as if with a plane) "*The machine shaved off fine layers from the piece of wood*"
- [S: \(v\) plane, skim](#) (travel on the surface of water)
- [S: \(v\) plane](#) (make even or smooth, with or as with a carpenter's plane) "*plane the top of the door*"

Adjective

- [S: \(adj\) flat, level, plane](#) (having a surface without slope, tilt in which no part is higher or lower than another) "*a flat desk*"; "*acres of level farmland*"; "*a plane surface*"; "*skirts sewn with fine flat seams*"

Word Sense Disambiguation (WSD): variants

Definition:

Lexical Sample task

- Small pre-selected set of target words to be disambiguated
- And inventory of senses for each word

All-words task

- Every word in an entire text to be disambiguated
- A lexicon with senses for each word
- Sort of like part-of-speech tagging
- Except each lemma has its own tagset



Word Sense Disambiguation (WSD)

Supervised

Unsupervised

Minimally supervised

Neural WSD

Recent Trends in Word Sense Disambiguation: A Survey

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Abstract

Word Sense Disambiguation (WSD) aims at making explicit the semantics of a word in context by identifying the most suitable meaning from a predefined sense inventory. Recent breakthroughs in representation learning have fueled intensive WSD research, resulting in considerable performance improvements, breaching the 80% glass ceiling set by the inter-annotator agreement. In this survey, we provide an extensive overview of current advances in WSD, describing the state of the art in terms of i) resources for the task, i.e., sense inventories and reference datasets for training and testing, as well as ii) automatic disambiguation approaches, detailing their peculiarities, strengths and weaknesses. Finally, we highlight the current limitations of the task itself, but also point out recent trends that could help expand the scope and applicability of WSD, setting up new promising directions for the future.

1 Introduction

Word Sense Disambiguation (WSD) is a historical task in Natural Language Processing (NLP) and Artificial Intelligence (AI) which, in its essence, dates back to Weaver [1949], who recognized the problem of polysemous words in the context of Machine Translation. Even today, word polysemy remains one of the most challenging and pervasive linguistic phenomena in NLP. For example, the ambiguous word *bass* refers to two completely disjoint classes of objects in the following sentences: i) “I can hear *bass* sounds”, ii) “They like *grilled bass*”. No approach has been sought since to tackle

and adverbs), as these are the words carrying most of a sentence’s meaning. In WSD, the sense inventory for a language can be very large, i.e., in the order of hundreds of thousands of concepts, but also very sparse, in that each *lexeme*¹ is associated with only a small subset of the sense inventory.

Predefined inventories define the output space for most varieties of past and modern approaches. These exist in many flavors, ranging from purely supervised [Hadwinoto *et al.*, 2019; Bevilacqua and Navigli, 2019] to knowledge-based [Moro *et al.*, 2014; Agirre *et al.*, 2014; Scozzafava *et al.*, 2020], to hybrid supervised and knowledge-based approaches [Kumar *et al.*, 2019; Bevilacqua and Navigli, 2020; Blevins and Zettlemoyer, 2020; Conia and Navigli, 2021; Barba *et al.*, 2021]. Supervised models, today based on neural architectures, frame the task as a classification problem and take advantage of annotated data to learn the association between words² in context and senses. Knowledge-based approaches, instead, often employ graph algorithms on a semantic network, in which senses are connected through semantic relations and are described with definitions and usage examples. Their independence from labeled training data, however, comes at the expense of performing worse than supervised models [Pilehvar and Navigli, 2014; Raganato *et al.*, 2017a; Pasini *et al.*, 2021] which, benefiting from pretrained language models, can now also nimbly scale across different languages. Nonetheless, information in semantic networks, be it unstructured (e.g., definitions) or structured (e.g., relational information), still remains highly relevant. This is demonstrated by hybrid approaches, which, reporting the highest results in literature, are currently attested as the best solution [Barba *et al.*, 2021].

Considering the fast pace at which the field is moving, to-

<https://www.ijcai.org/proceedings/2021/593>

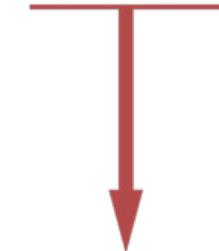


Entity Linking

From Wikipedia:

Entity linking, also referred to as named-entity linking (NEL), named-entity disambiguation (NED), named-entity recognition and disambiguation (NERD) or named-entity normalization (NEN) is the task of assigning a unique identity to entities (such as famous individuals, locations, or companies) mentioned in text.

“Paris is the capital of France”



wikipedia.org/wiki/Paris

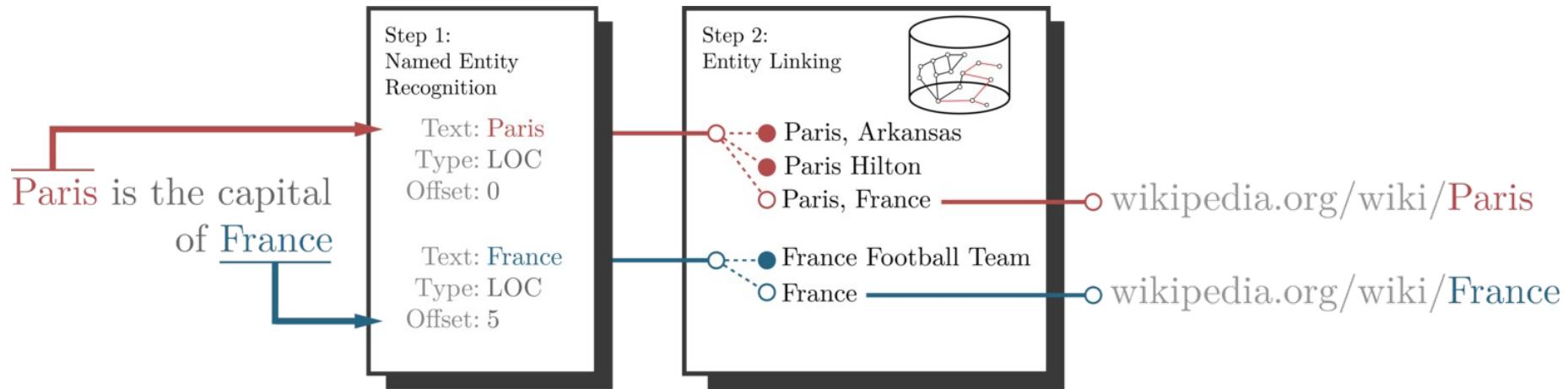


wikipedia.org/wiki/France

example of text
wikification



Entity Linking



From Wikipedia:

- **Text-based approaches**, which make use of textual features extracted from large text corpora (e.g. [Term frequency–Inverse document frequency](#) (Tf–Idf), word co-occurrence probabilities, etc...). [\[25\]](#)[\[16\]](#)
- **Graph-based approaches**, which exploit the structure of [knowledge graphs](#) to represent the context and the relation of entities. [\[3\]](#)[\[26\]](#)

Relation Extraction

“Paris is the capital of France”



capitalOf(Paris,France)

https://en.wikipedia.org/wiki/Relationship_extraction

Ontology Learning

From Wikipedia:

Ontology learning (**ontology extraction**, **ontology generation**, or **ontology acquisition**) is the automatic or semi-automatic creation of ontologies, including extracting the corresponding domain's terms and the relationships between the concepts that these terms represent from a corpus of natural language text, and encoding them with an ontology language for easy retrieval. As building ontologies manually is extremely labor-intensive and time-consuming, there is great motivation to automate the process.

Typically, the process starts by extracting terms and concepts or noun phrases from plain text using linguistic processors such as part-of-speech tagging and phrase chunking. Then statistical^[1] or symbolic^{[2][3]} techniques are used to extract relation signatures, often based on pattern-based^[4] or definition-based^[5] hypernym extraction techniques.

https://en.wikipedia.org/wiki/Ontology_learning



Volume 39, Issue 3

September 01 2013



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Abstract

- 1. Introduction
- 2. Related Work
- 3. The Taxonomy Learning Workflow
- 4. Evaluation
- 5. Conclusions
- Acknowledgments
- Notes

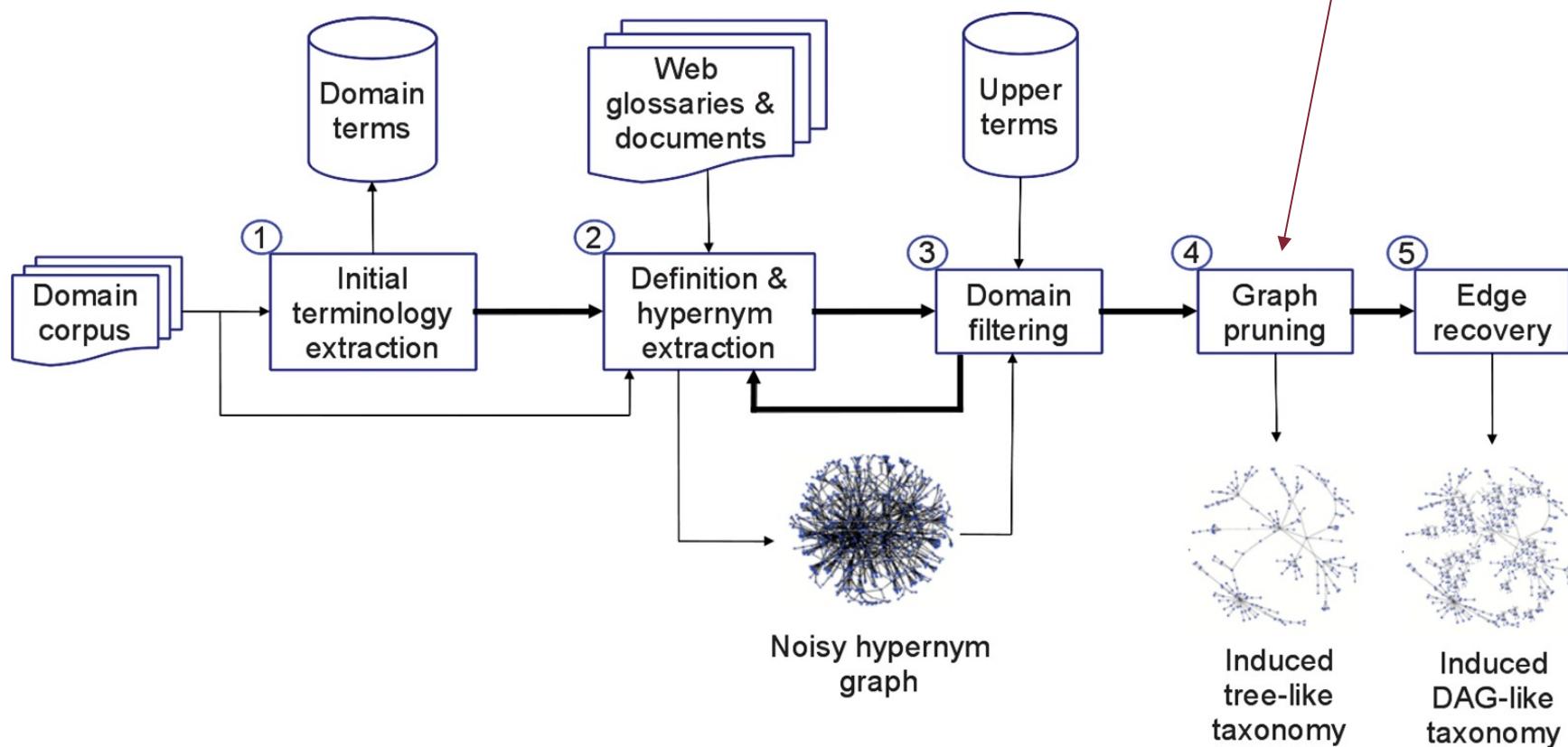
Abstract

In 2004 we published in this journal an article describing OntoLearn, one of the first systems to automatically induce a taxonomy from documents and Web sites. Since then, OntoLearn has continued to be an active area of research in our group and has become a reference work within the community. In this paper we describe our next-generation taxonomy learning methodology, which we name OntoLearn Reloaded. Unlike many taxonomy learning approaches in the literature, our novel algorithm learns both concepts and relations entirely from scratch via the automated extraction of terms, definitions, and hypernyms. This results in a very dense, cyclic and potentially disconnected hypernym graph. The algorithm then induces a taxonomy from this graph via optimal branching and a novel weighting policy. Our experiments show that we obtain high-quality results, both when building brand-new taxonomies and when reconstructing sub-hierarchies of existing taxonomies.

<https://direct.mit.edu/coli/article/39/3/665/1442/OntoLearn-Rerloaded-A-Graph-Based-Algorithm-for>

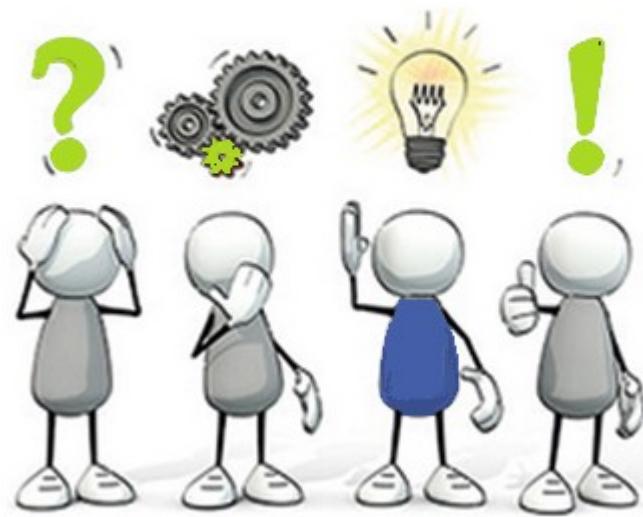
Ontolearn Reloaded

One of my current research interests



https://www.youtube.com/watch?v=-k3cOEoI_Dk

Q&A



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Word Sense Disambiguation as a Wordnets' Validation Method in Balkanet. In Proceedings of the Fourth International Conference on Language Resources and Evaluation (LREC'04), Lisbon, Portugal. European Language Resources Association (ELRA).
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Resources and References

[**Navigli&Ponzetto, 2010**] Roberto Navigli and Simone Paolo Ponzetto. 2010. BabelNet: Building a Very Large Multilingual Semantic Network. In *Proceedings of the 48th Annual Meeting of the Association for Computational Linguistics*, pages 216–225, Uppsala, Sweden. Association for Computational Linguistics. <https://aclanthology.org/P10-1023/>

[**Navigli&Ponzetto, 2012**] Roberto Navigli, Simone Paolo Ponzetto, BabelNet: The automatic construction, evaluation and application of a wide-coverage multilingual semantic network, *Artificial Intelligence*, Volume 193, 2012, Pages 217-250, ISSN 0004-3702, <https://doi.org/10.1016/j.artint.2012.07.001>

****Credits**

The slides of this part of the course are the result of a personal reworking of the slides and of the course material from different sources:

1. The NLP course of Prof. Roberto Navigli, Sapienza University of Rome
2. The NLP course of Prof. Simone Paolo Ponzetto, University of Mannheim, Germany
3. The NLP course of Prof. Chris Biemann, University of Hamburg, Germany
4. The NLP course of Prof. Dan Jurafsky, Stanford University, USA