Language Processing with Perl and Prolog Chapter 15: Lexical Semantics

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Words and Meaning

Referred to as lexical semantics:

- Classes of words: If it is hot, can it be cold?
- Definition What is a meal? What is table?
- Reasoning: The meal is on the table. Is it cold?

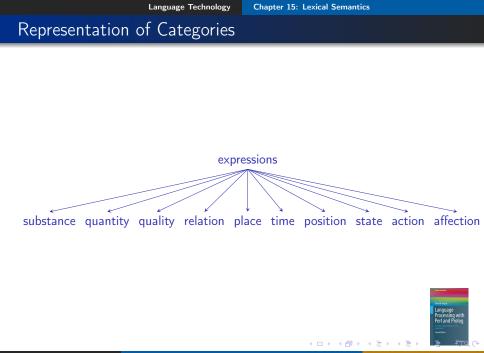
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Perl and Pro

Categories of Words

Expressions, which are in no way composite, signify substance, quantity, quality, relation, place, time, position, state, action, or affection. To sketch my meaning roughly, examples of substance are 'man' or 'the horse', of quantity, such terms as 'two cubits long' or 'three cubits long', of quality, such attributes as 'white', 'grammatical'. 'Double', 'half', 'greater', fall under the category of relation; 'in the market place', 'in the Lyceum', under that of place; 'yesterday', 'last year', under that of time. 'Lying', 'sitting', are terms indicating position, 'shod', 'armed', state; 'to lance', 'to cauterize', action; 'to be lanced', 'to be cauterized', affection.

Aristotle, Categories, IV. (trans. E. M. Edghill)



Classes

- Synonymy/Antonymy
- Polysemy
- Hyponyms/Hypernyms is_a(tree, plant), life form, entity
- Meronyms/Holonyms part_of(leg, table)
- Grammatical cases: [nominative]] broke [accusative the window] [ablative with a hammer]
- Semantic cases: [actor] broke [object the window] [instrument with a hammer]
- Case ambiguity (The window broke/ I broke the window)



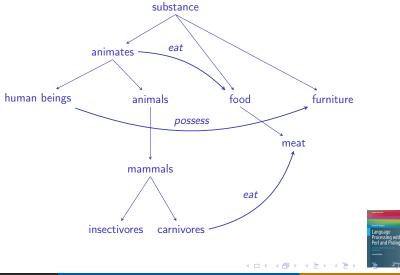
Lexical Database

```
%% is_a(?Word, ?Hypernym)
is_a(hedgehog, insectivore).
is_a(cat, feline).
is_a(feline, carnivore).
is_a(insectivore, mammal).
is_a(carnivore, mammal).
is_a(mammal, animal).
is_a(animal, animate_being).
```

hypernym(X, Y) :- is_a(X, Y). hypernym(X, Y) :- is_a(X, Z), hypernym(Z, Y).

Processing with Processing with Perl and Process International Sources

Semantic Networks



An Example: WordNet

Nouns hyponyms/hypernyms synonyms/antonyms meronyms Adjectives synonyms/antonyms relational fraternal -> brother Verbs Semantic domains (body function, change, communication, perception, contact, motion, creation, possession, competition, emotion, cognition, social interaction, weather) Synonymy, Antonymy: (rise/fall, ascent/descent, live/die) "Entailment": succeed/try, snore/sleep

Semantics and Reasoning

The caterpillar ate the hedgehog.

Representation:

 $\exists (X, Y), caterpillar(X) \land hedgehog(Y) \land ate(X, Y).$

Reasoning (inference): It is untrue because the query:

?- predator(X, hedgehog)
X = foxes, eagles, car drivers, ...

but no caterpillar.

Words are ambiguous: A same form may have more than one entry and sense.

The Oxford Advanced Learner's Dictionary (OLAD) lists five entries for bank:

- 1 noun, raised ground
- 2 verb, turn
- Inoun, organization
- verb, place money
- Inoun, row or series

and five senses for the first entry.

Short texts describing a word:

- A genus or superclass using a hypernym.
- Specific attributes to differentiate it from other members of the superclass. This part of the definition is called the *differentia specifica*.

bank (1.1): a land sloping up along each side of a canal or a river.

hedgehog: a small animal with stiff spines covering its back.

waiter: a person employed to serve customers at their table in a restaurant, etc.



Significance of the Sense

French	Geri	man	Danish
arbre	Ba	um	
	На	olz	Træ
bois			
forêt	Wa	ald	Skov
Fr	ench	We gwy	
	/ert		
L	bleu	gla	95
ł	gris		
		llw	yd



brun

Language Technology

Sense Tagging Using the Oxford Advanced Learner's Dictionary (OALD)

Sentence: The patron ordered a meal

Words	Definitions	Sense
The patron	Correct sense: A customer of a shop, restaurant,	1.2
	theater	
	Alternate sense: A person who gives money or sup-	1.1
	port to a person, an organization, a cause or an ac-	
	tivity	
ordered	Correct sense: To request somebody to bring food,	2.3
	drink, etc in a hotel, restaurant etc.	
	Alternate senses: To give an order to somebody	2.1
	To request somebody to supply or make goods, etc.	2.2
	To put something in order	Language Processing with
a meal	Correct sense: The food eaten on such occasion	Perl and Prolog
	Alternate sense: An occasion where food is eaten	T MARCO
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Identifying Senses

Semantic tagging looks like POS tagging: it assumes the sense of a word depends on its context.

We analyze the interaction between **bank** and market finance in a model where bankers gather information through monitoring. . .

Statistical techniques optimize a sequence of semantic tags. The context C of word w is defined as:

 $W_{-m}, W_{-m+1}, ..., W_{-1}, W, W_1, ..., W_{m-1}, W_m.$

If w has n senses, $s_1..s_n$, the optimal sense given C is defined as:

 $\hat{s} = \underset{s_i, 1 \leq i \leq n}{\operatorname{arg\,max}} P(s_i | C).$

Using Bayes' rule, we have:

$$\hat{s} = rgmax_{s_i,1 \le i \le n} P(s_i) P(C|s_i),$$

 $= \operatorname{arg\,max}_{s_i,1 \leq i \leq n} P(s_i) P(w_{-m}, w_{-m+1}, \dots, w_{-1}, w_1, \dots, w_{m-1}, w_m | s_i)$

Naïve Bayes

The Naïve Bayes classifier uses the bag-of-word approach. We replace

$$P(w_{-m}, w_{-m+1}, ..., w_{-1}, w_1, ..., w_{m-1}, w_m | s_i)$$

with the product of probabilities:

$$\prod_{j=-m,j\neq 0}^m P(w_j|s_i).$$

SemCor is a sense-annotated corpus for English. Semisupervised and unsupervised algorithms



Using Dictionaries (Lesk and derived methods)

We analyze the interaction between **bank** and market **finance** in a model where bankers gather information through monitoring and screening

Maximally overlapping definitions (Oxford Advanced Learner's Dictionary, 1995):

- Bank:
 - Sense 1: The land sloping up along each side of a river or a canal; the ground near a river
 - Sense 3: An organization or a place that provides a financial service. Customers keep their **money** in the bank safely and it is paid out when needed by the means of cheques, etc.

• Finance:

Sense 1: The **money** used or needed to support an activity, project, etc; the management of **money**

Beyond Words: Predicates and Arguments

Dictionaries store information about how words combine with other words to form larger structures.

This information is called valence (cf. valence in chemistry)

In the *Oxford Advanced Learner's Dictionary*, **tell**, sense 1, has the valence patterns:

tell something (to somebody) / tell somebody (something) as in:

- I told a lie to him
- I told him a lie

Both have the same predicate-argument representation:

tell.01(Speaker: I, Utterance: a lie, Hearer: him)

Case Grammar

Verbs have semantic cases (or semantic roles):

- An Agent Instigator of the action (typically animate)
- An Instrument Cause of the event or object in causing the event (typically animate)
- A Dative Entity affected by the action (typically animate)
- A Factitive Object or being resulting from the event
- A Locative Place of the event
- A Source Place from which something moves,
- A Goal Place to which something moves,
- A Beneficiary Being on whose behalf the event occurred (typically animate)
- A Time Time at which the event occurred
- An Object Entity that is acted upon or that changes, the most general case.

Case Grammar: An Example

open(Object, {Agent}, {Instrument})

The door opened John opened the door The wind opened the door John opened the door with a chisel Object = door Object = door and Agent = John Object = door and Agent = wind Object = door, Agent = John, and Instrument = chisel



Parsing with Cases

The waiter brought the meal to the patron

Identify the verb bring and apply constraints:

Case	Туре		Value
Agentive	Animate	(Obligatory)	The waiter
Objective (or theme)		(Obligatory)	the meal
Dative	Animate	(Optional)	the patron
Time		(Obligatory)	past



FrameNet

In 1968, Fillmore wrote an oft cited paper on case grammars. Later, he started the FrameNet project: http://framenet.icsi.berkeley.edu/ Framenet is an extensive lexical database itemizing the case (or frame) properties of English verbs. In FrameNet, Fillmore no longer uses universal cases but a set of frames – predicate argument structures – where each frame is specific to a class of words.

The Impact Frame

Impact:

bang.v, bump.v, clang.v, clunk.v, collide.v, collision.n, crash.v, crash.n, crunch.v, glancing.a, graze.v, hit.v, hit.n, impact.v, impact.n, plop.v, plough.v, plunk.v, run.v, slam.v, slap.v, smack.v, smash.v, strike.v, thud.v, thump.v

Frame elements:

cause, force, impactee, impactor, impactors, manner, place, result, speed, sub_location, time.

Protection Protection Processing with Perl and Prolog Processing with Perl and Prolog Protection Pr

The Revenge Frame

15 lexical units (verb, nouns, adjectives):

avenge.v, avenger.n, get back (at).v, get_even.v, retaliate.v, retaliation.n, retribution.n, retributive.a, retributory.a, revenge.n, revenge.v, revengeful.a, revenger.n, vengeance.n, vengeful.a, and vindictive.a.

Five frame elements (FE):

Avenger, Punishment, Offender, Injury, and Injured_party.

The lexical unit in a sentence is called the target.

Annotation

- [<*Avenger>* His brothers] avenged [<*Injured_party>* him].
- With this, [<Avenger> El Cid] at once avenged [<Injury> the death of his son].
- [<Avenger> Hook] tries to avenge [<Injured_party> himself] [<Offender> on Peter Pan] [<Punishment> by becoming a second and better father].

FrameNet uses three annotation levels: Frame elements, Phrase types (categories), and grammatical functions. GFs are specific to the target's part-of-speech (i.e. verbs, adjectives, prepositions, and nouns). For the verbs, three GFs: Subject (Ext), Object (Obj), Complement (Dep), and Modifier (Mod), i.e. modifying adverbs ended by *-ly* or indicating manner

manner

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The Valence Pattern

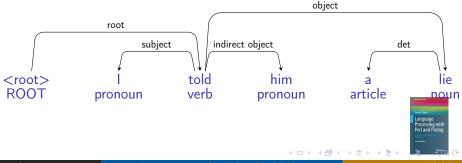
Sent. 1	avenge	FE	Avenger	Injured_party		
		PT	NP	NP		
		GF	Ext	Object		
Sent. 2	avenge	FE	Avenger	Injury		
		ΡT	NP	NP		
		GF	Ext	Obj		
Sent. 3	avenge	FE	Avenger	Injured_party	Offender	Punishment
		ΡT	NP	NP	PP	PPing
		GF	Ext	Obj	Comp	Comp
				< • • •	<₽> <≥>	I answer I anguage Processing with Peti and Prolog Weren Market

Automatic Frame-semantic Analysis (Johansson, 2008)

Given a sentence:

I told him a lie

and a target word – **tell** –, find the semantic arguments. In Propbank, the possible arguments of **tell.01** are *speaker* (Arg0), *utterance* (Arg1), and *hearer* (Arg2) Input: a syntax tree:



Classification of Semantic Arguments (Johansson, 2008)

Two steps:

- Find the arguments,
- Determine the role (name) of each argument

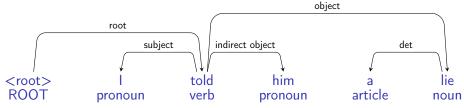
The identification of semantic arguments can be modeled as a statistical classification problem.

What features are useful for this task? Examples:

- Grammatical function: subject, object, ...
- Voice: I told a lie / I was told a lie
- Semantic classes: I told him / the note told him
- Semantic class usually not available: use word instead

Feature Extraction (Johansson, 2008)

Given a dependency tree:



We select the three dependents of *told* and we extract features to determine if it is a semantic argument and its name.

Grammatical function		0
Subject	Active	speaker (Arg0)
Indirect object	Active	hearer (Arg2)
Direct object	Active	utterance (Arg1)
	Subject Indirect object	Indirect object Active

Propbank

Semantic analysis often uses Propbank instead of Framenet because of Propbank's larger annotated corpus

CoNLL 2008 and 2009 used Propbank for their evaluation of semantic parsers.

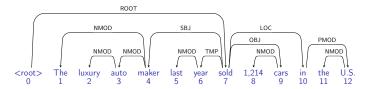
CoNLL annotation format of the sentence:

The luxury auto maker last year sold 1,214 cars in the U.S.

ID	Form	Lemma	PLemma	POS	PPOS	Feats	PFeats	Head	PHead	Deprel	PDeprel	FillPred	Sense	APred1	APred2
1	The	the	the	DT	DT			4	4	NMOD	NMOD				
2	luxury	luxury	luxury	NN	NN			3	3	NMOD	NMOD			Ā1	
3	auto	auto	auto	NN	NN	_	_	4	4	NMOD	NMOD	_	_	A1	-
4	maker	maker	maker	NN	NN	-	_	7	7	SBJ	SBJ	Ŷ	maker.01	A0	Ā0
5	last	last	last	IJ	JJ	-	_	6	6	NMOD	NMOD				
6	year	year	year	NN	NN	-	_	7	7	TMP	TMP	-	-	-	AM-TMP
7	sold	sell	sell	VBD	VBD	-	_	0	0	ROOT	ROOT	Ŷ	sell.01	-	
8	1,214	1,214	1,214	CD	CD			9	9	NMOD	NMOD				
9	cars	car	car	NNS	NNS			7	7	OBJ	OBJ				Ā1
10	in	in	in	IN	IN			7	7	LOC	LOC				AM-LOC
11	the	the	the	DT	DT			12	12	NMOD	NMOD				
12	U.S.	u.s.	u.s.	NNP	NNP			10	10	PMOD	PMOD				Language
														-	Processing with Perl and Prolog
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Visualizing Dependencies

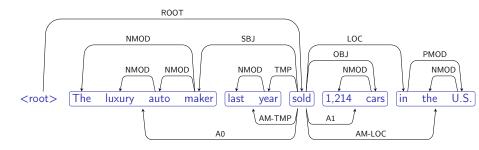
Syntactic dependencies:



Semantic dependencies (predicate-argument structures):

-Root-	The 1 the the DT	luxury 2 luxury luxury	auto auto auto	maker 4 maker maker	last 5 last last	vear year year year	sold 7 sell sell	AM-LA A1 1,214 1,214 1,214	cars 9 car car	in 10 in	the 11 the the	U.S. 12 u.s. u.s.
	DT	NN NN	NN NN	NN NN	1) 1)	NN NN	VBD VBD	CD CD	NNS NNS	IN IN	DT DT	NNP NNP
	_	_	_	- (maker 01)	_	_	(sell.01)	_	_	_	_	_

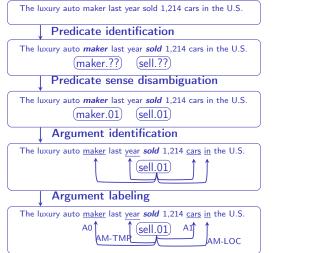
Alternative Visualization



	The	luxury	auto	maker	last	year	sold	1,214	cars	in	the	U.S.
maker.01		A1		A0								
sell.01		A0			AM	TMP		A	1		AM-L	OC
]	•		

Parsing Pipeline





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Perl and Prolo

Parsing Components

Almost all the semantic parsers (or semantic role labelers) start with a parsing step: either dependencies or constituents. The semantic parser consists of a sequence of classifiers. Logistic regression is among the best classifiers. Each classifier uses a set of features extracted from the previous steps.



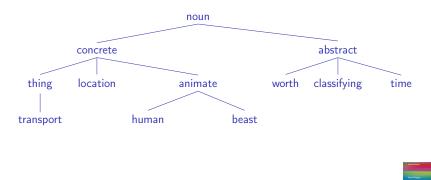
Features for the Predicate Identification

Features used by Johansson and Nugues (2008) and values for *sold* in *The luxury auto maker last year sold* 1,214 *cars in the U.S.*

Feature	Value
PredForm	sold
PredLemma	sell
PredHeadForm	ROOT
PredHeadPOS	ROOT
PredDeprel	ROOT
ChildFormSet	{maker, year, cars, in}
ChildPOSSet	{NN, NNS, IN}
ChildDepSet	{SBJ, TMP, OBJ, LOC}
DepSubcat	SBJ+TMP+OBJ+LOC
ChildFormDepSet	{maker+SBJ, year+TMP, cars+OBJ, in+LO
ChildPOSDepSet	{NN+SBJ, NN+TMP, NNS+OBJ, IN+LOC
	Applied and a second seco

EVAR

EVAR is a German project that aims at providing information on trains



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Processing with Perl and Prolo

EVAR's Case Grammar

• fahren1.1 (*The train is going from Hamburg to Munich*)

- Instrument: noun group (nominative), Transport, obligatory
- Source: prepositional group (Origin), Location, optional
- Goal: prepositional group (Direction), Location, optional

2 fahren1.2 (I am going by train from Hamburg to Munich)

- Agent: noun group (nominative), Animate, obligatory
- Instrument: prepositional group (prep = mit), Transport, optional
- Source: prepositional group (Origin), Location, optional
- Goal: prepositional group (Direction), Location, optional

S Abfahrt1.1 (The departure of the train at Hamburg for Munich)

- Object: noun group (genitive), Transport, optional
- Location: prepositional group (Place), Location, optional
- Time: prepositional group (Moment), Time, optional

Application: Carsim

Identify the events (actions) and the semantic relations related to car accidents.

In Framenet, the **Impact** class consists of 38 verbs or nouns with the roles: **Impactor**, **Impactee**, **Impactees**

[<Impactor> The rock] HIT [<Impactee> the sand] with a thump Source: http://framenet.icsi.berkeley.edu/ In Carsim:

i ett äldreboende] [_{LOC} på Alvägen] [_{LOC} i Enebyberg] [_{LOC}

[ACTOR En personbil] körde [TIME vid femtiden] [TIME på torsdagseftermiddagen] in [VICTIM i ett radhus] [LOC

norr om Stockholm].