

Introduction

Semantic classes and polysemy

Experiment 1: independent classes

Experiment 2: multi-label classification

Discussion

Research agenda

# Modeling regular polysemy: A study in the semantic classification of Catalan adjectives

Gemma Boleda

Universitat Pompeu Fabra

September 6, 2011

# Outline

- 1 Introduction
- 2 Semantic classes and polysemy
- 3 Experiment 1: independent classes
- 4 Experiment 2: multi-label classification
- 5 Discussion
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## Overview

Automatic acquisition of **semantic classes for adjectives**  
[Boleda et al., 2007, Boleda et al., prep]

**Given:** classification, set of adjectives, corpus

**Task:** infer the class for each adjective in the set

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- we have to build these as we go along
  - exploratory nature of the experiments
- two classifications tested
  - experiments 1 and 2

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- some adjectives are **polysemous** → have more than one sense → belong to more than one class
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# Hypotheses and contribution

## Lexical Acquisition

- 1 there is a stable relationship between semantic properties and other linguistic properties
  - Distributional Hypothesis [Harris, 1968]
- 2 linguistic properties can be modeled using observable cues in corpora
- 3 → observable cues can be used to induce semantic properties

# Hypotheses and contribution

## This study

- 1 it is possible to use observable cues to
  - induce semantic classes for adjectives
  - identify polysemous adjectives
- 2 computational methodology can give feedback to theoretical questions

# Hypotheses and contribution

## This study

- 1 it is possible to use observable cues to
  - induce semantic classes for adjectives
    - most work on lexical acquisition on verbs (vs. adjectives)
    - and English (vs. Catalan)
  - identify polysemous adjectives
- 2 computational methodology can give feedback to theoretical questions

# Hypotheses and contribution

## This study

- 1 it is possible to use observable cues to
  - induce semantic classes for adjectives
  - identify polysemous adjectives
    - polysemy largely ignored in related work on lexical acquisition
    - regular polysemy: studied on a theoretical level, not in empirical approaches to computational semantics
- 2 computational methodology can give feedback to theoretical questions

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# Semantic classification

- insights from descriptive grammar and formal semantics

**Qualitative adjectives** denote attributes or properties of objects.

Examples: ample, autònom  
'wide', 'autonomous'

**Intensional adjectives** denote second order properties.

Examples: presumpte, antic  
'alleged', 'former'

**Relational adjectives** denote a relationship to an object.

Examples: pulmonar, botànic  
'pulmonary', 'botanical'

- semantic classes
- correlate with other linguistic properties

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## Example property: predicativity

qualitative (1) predicative	intensional (2) non-predicative	relational (3) marginally predicative
--------------------------------	------------------------------------	------------------------------------------

(1) el carrer és ample  
the street is wide

(2) #l' assassí és presumpte  
the murderer is alleged

(3) ?la malaltia és pulmonar  
the disease is pulmonary

# Polysemy

- Polysemy cutting across two classes: relational (4a) and qualitative (4b):

- (4) a. la recuperació econòmica  
the recovery economic  
'the economic recovery'
- b. els pantalons econòmics  
the trousers economic  
'the cheap trousers'

- (5) a. ?la recuperació és econòmica  
the recovery is economic  
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- b. els pantalons són econòmics  
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in each sense, the adjective's behaviour corresponds to that of the relevant class

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## Regular polysemy [Apresjan, 1974, Copestake and Briscoe, 1995]

- same type of polysemy for a range of adjectives
  - (6) a. reunió familiar / cara familiar  
meeting familiar / face familiar  
'family meeting / familiar face'
  - b. problema amorós / noi amorós  
problem love<sub>SUFFIX</sub> / boy love<sub>SUFFIX</sub>  
'love problem / lovely boy'
- in general:  
relation to **object** → salient property of the **object**
- we only consider class-related polysemy

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# Experiment 1: Motivation

- classification based on linguistic literature
- does it account for the semantics of a broad range of adjectives?
- polysemous adjectives should exhibit a different profile than monosemous adjectives
- is this behaviour distinct enough to identify polysemous classes?

## Experiment 1: Motivation

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## Material and method (I)

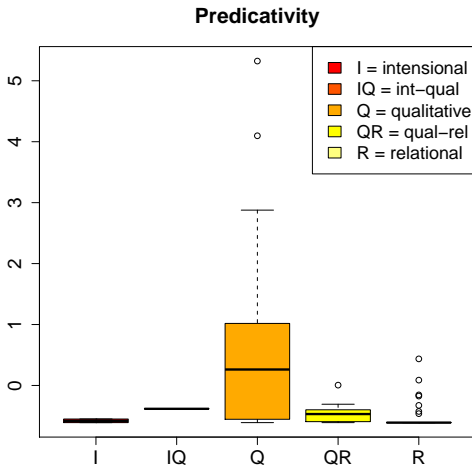
- CTILC corpus (*Institut d'Estudis Catalans*):
  - 14.5 million words, written, formal texts
  - manually lemmatised and POS-tagged
  - automatically shallow-parsed (noise)
- adjective database [Sanromà and Boleda, 2010]:
  - almost 2,300 lemmata from CTILC corpus
  - morphological information manually coded

## Material and method (II)

- Gold Standard: 101 lemmata, random choice **except for intensional class**
  - 4 judges,  $\kappa$  0.54-0.64
  - for each adjective, choose semantic class
  - target classes:
    - basic classes: qualitative (Q), relational (R)
    - polysemous “class”: qualitative-relational (QR)
- technique: clustering,  $k$ -means
  - 3,521 objects (freq > 10)
- features:
  - **theoretically motivated**: predicativity, ... (6)
  - **POS**: POS unigrams; 2 words left and right of target (36)

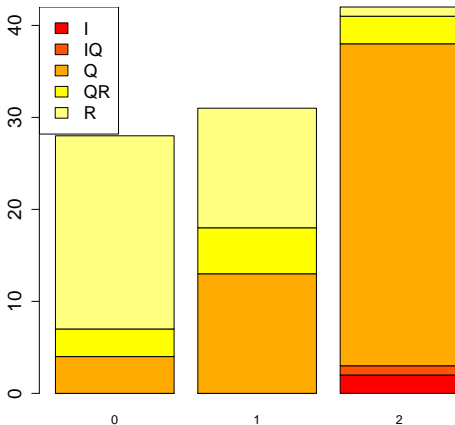


# Feature example: value distribution across classes

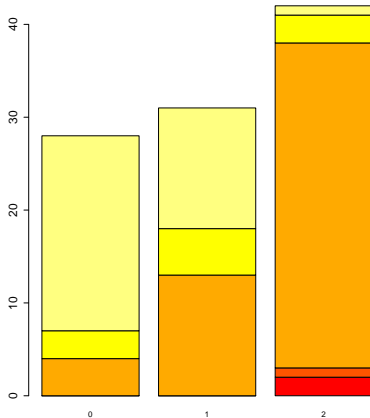
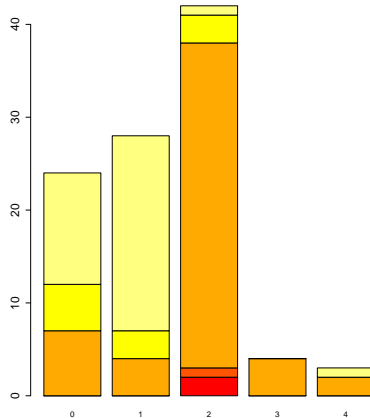


# Results

### Theoretical features, 3 clusters

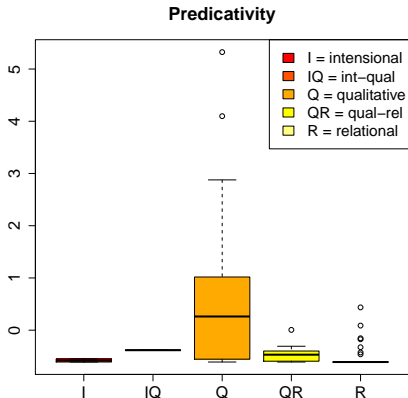


# Results

**A: 3 clusters****B: 5 clusters**

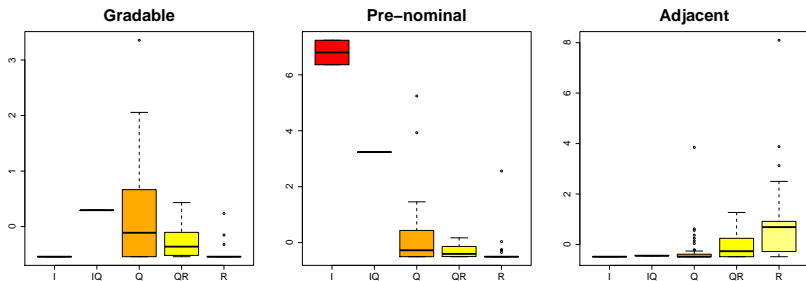
# Discussion: polysemy

- approach to polysemy is clearly wrong



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## Discussion: polysemy

- polysemous adjectives do not have a homogeneous, differentiated profile
- most adjectives are used predominantly in one of their senses, corresponding to one of the classes
  - **irònic** ('ironic'): qualitative-relational.
    - mainly used as qualitative in the corpus
    - systematically assigned to the qualitative cluster
  - **militar** ('military'): qualitative-relational.
    - mainly used as relational in the corpus
    - systematically assigned to the relational cluster

## Discussion: classification

- “mixed” cluster: group of problematic adjectives identified in error analysis:
  - indicador, parlant, protector, salvador, ...
  - ‘indicating’, ‘speaking’, ‘protecting’, ‘saviour’, ...
- these adjectives do not fit into the classification
- create new class

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# Modified classification

Qualitative adjectives

Relational adjectives

Event-related adjectives denote a relationship to an event.

Examples: protector, variable  
'protecting', 'variable'

- relationship with morphology
  - qualitative      event      relational
  - non-derived    deverbal    denominal
- supported by Ontological Semantics  
[Raskin and Nirenburg, 1998]

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## Experiment 2: Motivation

Experiment 1 shows that

- polysemous adjectives *do* exhibit a different profile from monosemous adjectives
  - ... but it is *not* distinct enough to identify polysemous classes
  - polysemy = membership in more than one class
- multi-label classification
- a lemma can belong to more than one target class
  - look for properties of each of the classes

## Material: Gold Standard

- same corpus and database as in Experiment 1
- Gold Standard: 210 lemmata
- stratified sampling approach
  - frequency, morphology
- large-scale manual annotation experiment
  - task: choose **one or more** pseudo-dictionary definitions
  - administered via Web
  - 322 naive subjects
  - does not yield reliable classification ( $\kappa$  0.31-0.45)
- Gold Standard classification: committee of 3 experts
  - agreement subjects-experts:  $p_o$  0.68,  $\kappa$  **0.55**

## Method

- classifiers: Decision Trees (flat), ensemble classifiers
- features:

Type	Explanation	# F.
morph	morphological properties ex.: <i>suffix</i>	2
func	syntactic function ex.: <i>predicate in copular sentence</i>	4
uni	uni-gram distribution ex.: <i>-1noun</i>	24
bi	bi-gram distribution ex.: <i>-1noun+1adj</i>	50
theor	distributional cues of theoretical properties ex.: <i>gradable</i>	18

Table: Experiment 2: features.

# Procedure

## Standard procedure for multi-label classification

1. binary decision
2. merge classifications
  - **economic:**  
qualitative    relational    event    merged  
yes            yes            no            qualitative-relational (QR)
  - rationale: if an adjective is polysemous, it will exhibit properties of each class it belongs to

## Evaluation

- 10 run, 10-fold cross-validation
- baseline: most frequent class

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## Results

<b>Classifier</b>	<b>Accuracy</b>
<i>baseline</i>	$51.0 \pm 0.0$
best flat	$62.5 \pm 2.5$
Att. Bagg. <sub>FS,bin,i=100</sub>	<b><math>69.1 \pm 1.0</math></b>
<i>Human agreement</i>	<b>68</b>

Table: Experiment 2: summary of results.

## Error analysis

		Best classifier					Total	
		Q	E	R	QR	QE		ER
Experts	Q	90	4	2	3	8	0	107
	E	10	17	0	1	6	3	37
	R	4	0	20	4	0	2	30
	QR	5	0	4	13	0	1	23
	QE	1	1	0	0	5	0	7
	ER	0	0	2	1	0	3	6
Total		110	22	28	22	19	9	210

- overgenerated polysemous adjectives: 26
  - undergenerated polysemous adjectives: 13

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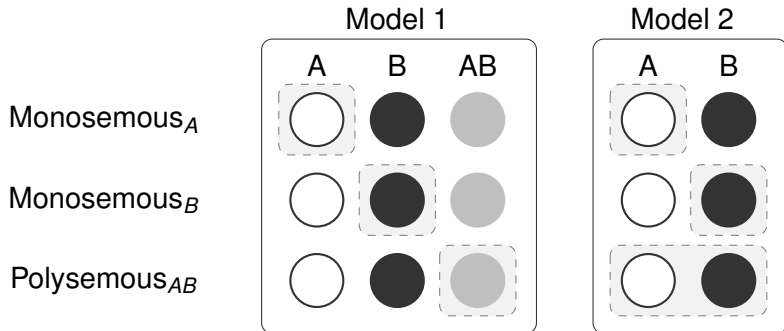
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## Discussion: Two models for regular polysemy



# Conclusion

## This study

- 1 it is possible to use observable cues to
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# Conclusion

## This study

- 1 it is possible to use observable cues to
  - induce semantic classes for adjectives
    - relationship between observable cues and semantic properties
    - explored morphology-semantics and syntax-semantics interfaces
    - roadblock: human agreement → need to improve theory
  - identify polysemous adjectives
- 2 computational methodology can give feedback to theoretical questions

# Conclusion

## This study

- 1 it is possible to use observable cues to
  - induce semantic classes for adjectives
  - identify polysemous adjectives
    - polysemous adjectives exhibit “hybrid” behaviour
    - tested two models of regular polysemy
    - need to model *both* similarities and differences with respect to basic classes
- 2 computational methodology can give feedback to theoretical questions



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- 2 computational methodology can give feedback to theoretical questions
  - random sampling: emergence of “nonprototypically nonprototypical” adjectives
    - Gold Standards: medium-sized datasets
  - feature representation: empirical properties
  - Machine Learning: evaluation of different models

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# Overall research question

How do *languages* encode *meaning*?

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## Overall research question

How do **words** and **word combinations** encode **meaning**?

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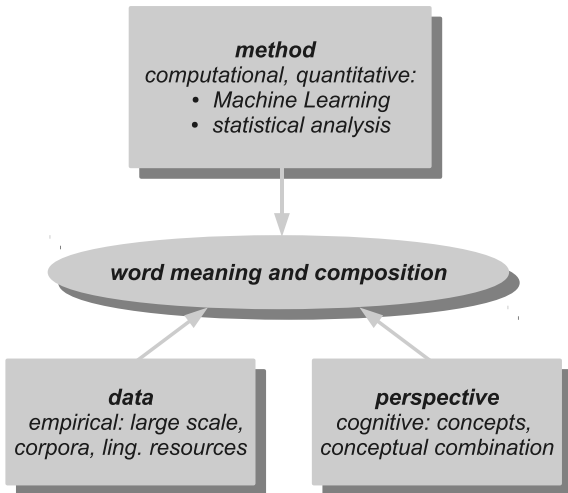
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# Overall research question

How do **words** and **word combinations** encode meaning?

→ **empirical computational lexical semantics**

# Approach: computational lexical semantics

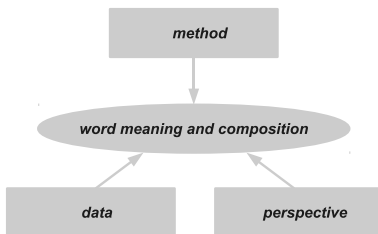


# Current research

## Goal

computationally model the interpretation processes that take place in semantic composition

→ distributional/vector-space models





# Question 1

- While words may have different meanings, when used in a given context not all of them are relevant.
- **Question 1: How does combining two words affect the interpretation of each word?**
  - colour adjectives: *red dress* vs. *red wine*.
    - the modifier is altered depending on the head noun.
- can distributional models account for the different meanings of the modifiers depending on the head noun?

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## Question 2

- Often, the interpretation of a complex expression is richer than what can be inferred from the meaning of its parts alone.
- **Question 2: How does combining two words yield the interpretation of a complex expression?**
  - noun-noun compounds: *dog magazine* [Murphy, 2002]  
→ magazine ABOUT dogs – and more!
  - can distributional models account for the relation between a head and its modifier?

## Question 2

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An applied ontological semantic microtheory of adjective meaning for natural language processing.

*Machine Translation*, 13(2-3):135–227.



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The database of catalan adjectives.

*In Proceedings of the Seventh conference on International Language Resources and Evaluation (LREC'10),*  
Valletta, Malta. European Language Resources Association (ELRA).

## Material for Experiment 2: Gold Standard

<b>class</b>	<b>#adjs.</b>	<b>Examples</b>
qualitative	107	<b>ample</b> 'wide'
event	37	<b>revelador</b> 'revealing'
relational	30	<b>pulmonar</b> , 'pulmonary'
qual-ev	7	<b>cridaner</b> 'vociferous/loud-coloured'
qual-rel	23	<b>amorós</b> 'affectionate/related to love'
ev-rel	6	<b>docent</b> 'teaching/related to teachers or the teaching task'



## Results for Experiment 2: linguistically principled EC

	A: Per-class accuracy			B: Overall accuracy	
	Qualit.	Event	Relat.	Full	Partial
best flat ( <i>all</i> )	<b>75.5</b> ±9.0	<b>86.5</b> ±6.4	86.0±6.5	62.5±2.5	<b>87.6</b> ±2.5
3 best feat. sets	72.9±1.4	<b>88.2</b> ±1.3	85.4±0.6	61.8±1.7	86.7±0.8
4 best feat. sets	74.6±2.0	<b>86.5</b> ±1.5	88.1±1.2	63.0±2.4	87.4±1.8
5 best feat. sets	75.2±2.7	86.4±1.4	<b>90.8</b> ±1.2	<b>64.8*</b> ±2.6	<b>89.5*</b> ±1.5
7 best feat. sets	75.3±2.3	82.7±1.8	<b>90.8</b> ±1.1	<b>64.0</b> ±1.5	85.9±1.3
all 9 feat. sets	<b>75.4</b> ±1.7	80.9±2.1	90.1±1.2	62.4±2.2	84.4±1.3

## Results for Experiment 2: other ensemble classifiers

	A: Per-class accuracy			B: Overall accuracy	
	Qualit.	Event	Relat.	Full	Partial
best flat ( <i>all</i> )	75.5±9.0	<b>86.5</b> ±6.4	86.0±6.5	62.5±2.5	87.6±2.5
best ling. EC (*)	75.2±2.7	<b>86.4</b> ±1.4	<b>90.8</b> ±1.2	64.8*±2.6	<b>89.5*</b> ±1.5
adaboost	<b>82.0*</b> ±8.6	85.6±7.1	88.0±6.7	66.0*±1.9	<b>89.9*</b> ±1.3
A. B. <sub>FS,bin</sub> , <i>i</i> =3	76.0±9.4	84.0±7.0	88.3±7.2	64.0±2.5	86.7±2.0
A. B. <sub>FS,bin</sub> , <i>i</i> =4	75.9±9.2	84.7±7.3	89.1±6.9	64.5±1.5	86.6±1.1
A. B. <sub>FS,bin</sub> , <i>i</i> =5	77.0±8.7	85.8±7.1	89.0±6.5	<b>66.3*</b> ±1.1	87.0±1.5
A. B. <sub>FS,bin</sub> , <i>i</i> =100	<b>81.0</b> ±8.8	86.1±6.9	<b>90.1*</b> ±5.3	<b>69.1***</b> ±1.0	89.0±1.0
<i>Human agreement</i>	-	-	-	68	85

## Variation in object-object modification

- (7) a. world war
- b. John's book
- c. agreement by France
- d. psychological evidence