

Towards Word Sense Disambiguation by Reasoning

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- 1 Introduction
- 2 New WSD Proposal
- 3 Experimentation
- 4 Conclusions and Ongoing Work

Word Sense Disambiguation (WSD) (Agirre and Edmonds, 2007)

- Natural Language Processing (NLP) research line
- Choosing the intended sense of a word in a context
- Example:
 - ▶ *turtle* is either a “*a sweater or jersey with a high close-fitting collar*” or “*any of various aquatic and land reptiles having a bony shell and flipper-like limbs for swimming*”
 - ▶ Contexts:
 - “*The black fleece is soft as cotton candy, and is second on my preferred *turtle* list*”
 - “*Some *turtles* lay eggs in the sand and leave them to hatch on their own. The young *turtles* make their way to the top of the sand and scramble to the water while trying to avoid predators*”

Knowledge Resources I

- **BLESS** data (Baroni-Lenci Evaluation of Semantic Similarity) (Baroni and Lenci, 2011)
 - ▶ Designed for the differentiation between semantic relations and compositionality
 - ▶ It includes 200 concrete nouns (*targets*) (100 animate and 100 inanimate nouns) from different classes (e.g., *tools*, *clothing*, *vehicles*, *animals*, etc.)
 - ▶ Each target is associated with a set of other words (nouns, verbs or adjectives) via:
 - *hyperonymy*
 - *attribute*
 - *cohyponymy*
 - *event*
 - *meronymy*
 - *random*
 - ▶ Example:
 - *turtles* are *amphibians* (hyperonymy)
 - *turtles* and *frogs* are coordinate (cohyponymy)
 - *turtles* have *legs* (meronymy)
 - *turtles* are *slow* (attribute)
 - *turtles* walk (event)

Knowledge Resources II

- **WordNet** (Fellbaum, 1998) is a large lexical database of English
- Nouns, verbs, adjectives and adverbs are grouped into sets of cognitive synonyms (*synsets*): lexical concepts that are synonyms
 - ▶ Example: $turtleneck_n^1$, $turtle_n^1$, $polo-neck_n^1$
- Synsets can be identified by its *offset*
 - ▶ Example: $turtle_n^1$ is identified by 04502197-n
- Synsets are related among them: hyponymy, meronymy, ...

Knowledge Resources III

- **Adimen-SUMO** (Álvarez et al., 2012) is a FOL ontology about common-sense knowledge
- It has been obtained by applying a reengineering process to SUMO (Niles and Pease, 2001)
 - ▶ With the help of ATPs (*Automated Theorem Provers*)
 - ▶ Around an 88% of the *core* of SUMO (top and middle levels) is translated
 - ▶ Domain ontologies are not used (by now)
- The resulting ontology can be used in tasks that involve reasoning with commonsense knowledge
- Example:

```
(=>
  ($instance ?LIMB Limb)
  (exists (?VERTEBRATE)
    (and
      ($instance ?VERTEBRATE Vertebrate))
      (properPart ?LIMB ?VERTEBRATE))))
```

Knowledge Resources IV

- **The mapping between WordNet and SUMO** (Niles and Pease, 2003)
- It connects synsets of WordNet to terms of SUMO using 3 relations:
 - ▶ *equivalence* (=)
 - ▶ *subsumption* (+)
 - ▶ *instance* (@)
- Some examples:

$\langle \text{turtle}_n^1 \rangle$:	$[\text{Clothing}_c+]$
$\langle \text{turtle}_n^2 \rangle$:	$[\text{Reptile}_c+]$
$\langle \text{clothing}_n^1 \rangle$:	$[\text{Clothing}_c=]$
$\langle \text{Snake}_n^4 \rangle$:	$[\text{AstronomicalBody}_c@]$

Snake_n^4 is "a long faint constellation in the southern hemisphere near the equator stretching between Virgo and Cancer"

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New WSD Proposal I

- Our objective is to demonstrate the practical capabilities of Adimen-SUMO for NLP tasks using Vampire (Kovács and Voronkov, 2013)
- On the basis of our black-box testing method (Álvarez et al., 2015, 2017b)
 - ▶ Adaptation of the methodology for the design and evaluation of ontologies introduced in (Grüniger and Fox, 1995)
 - ▶ Based on the use of **Competency Questions** (CQs):
 - Problems that an ontology is expected to answer
 - ▶ CQs are automatically created on the basis of few **Question Patterns** (QPs) by exploiting WordNet and its mapping into SUMO
 - ▶ Classification of (dual) problems (truth- and falsity-tests) by means of ATPs:
 - *Passing*: the ATPs are able to demonstrate a truth-test
 - *Non-passing*: the ATPs are able to demonstrate a falsity-test
 - *Unknown*: the ATPs produce no answer within a time limit

New WSD Proposal II

- We consider all the synsets to which words belong
 - ▶ Targets and related words
- We apply several predefined QPs to the resulting synset pairs
- For each target, we choose the most likely synset:
 - ▶ According to the difference between passing and non-passing CQs
- Example: *turtles* (2 synsets) *have* (meronymy) *legs* (9 synsets) → in total 18 synset pairs for this word pair
 - ▶ *turtle*_n² “any of various aquatic and land reptiles having a bony shell and flipper-like limbs for swimming”, *Reptile*_c+
 - ▶ *leg*_n¹ “a structure in animals that is similar to a human leg and used for locomotion”, *Limb*_c+
 - ▶ Resulting CQ:

```
(exists (?X ?Y)
  (and
    ($instance ?X Reptile)
    ($instance ?Y Limb)
    (properPart ?Y ?X)))
```

New WSD Proposal III

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 - ▶ Targets and related words
- We apply several predefined QPs to the resulting synset pairs
- For each target, we choose the most likely synset:
 - ▶ According to the difference between passing and non-passing CQs
- Example: *turtles* (2 synsets) *have* (meronymy) *legs* (9 synsets) -> in total 18 synset pairs for this word pair
 - ▶ *turtle*_n¹ “a sweater or jersey with a high close-fitting collar”, *Clothing*_c+
 - ▶ *leg*_n³ “one of the supports for a piece of furniture”, *Artifact*_c+
 - ▶ Resulting CQ:

```
(exists (?X ?Y)
  (and
    ($instance ?X Clothing)
    ($instance ?Y Artifact)
    (properPart ?Y ?X)))
```

New WSD Proposal IV

- There are 14,400 BLESS pairs (excluding *random*)
- We obtain 386,891 synset pairs, from which 127,918 synset pairs yield to incorrect formulas
- From the remaining 258,973 synset pairs, the resulting CQs are:

Pattern	#1	#2	#3	#4	Total
<i>hyperonymy</i>	2,559	458	–	–	3,017
<i>cohyponymy</i>	6,243	1,147	1,128	204	8,722
<i>meronymy</i>	6,669	1,852	1,556	393	10,470
<i>attribute</i>	3,092	600	685	109	4,486
<i>event</i>	12,575	3,220	3,357	721	19,873
Total	–	–	–	–	46,568

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Summary

- Adimen-SUMO v2.6
 - ▶ Including the *Closed World Assumption* (CWA) (Reiter, 1978) applied to *\$subclass* and *\$disjoint*
- Vampire v4.2.2 in an Intel® Xeon® CPU E5-2640v3@2.60GHz
 - ▶ CASC portfolio
 - ▶ Resource limits: 300 seconds and 2GB of memory
- Raw results:

Pattern	#1		#2		#3		#4		Solved	
<i>hyperonymy</i>	+1,027	-944	+173	-157	-	-	-	-	1,201	39.81%
<i>cohyponymy</i>	+712	-364	+189	-23	+216	-14	+27	-4	1,549	17.76%
<i>meronymy</i>	+785	-6	+37	-0	+109	-0	+0	-0	937	8.95%
<i>attribute</i>	+90	-84	+8	-1	+4	-3	+0	-0	190	4.24%
<i>event</i>	+0	-0	+0	-0	+0	-0	+0	-0	0	0.00%
Total	-	-	-	-	-	-	-	-	3,877	8.33%

WSD Results and Comparison

- We have manually constructed a *gold standard* for the 200 targets

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- We have manually constructed a *gold standard* for the 200 targets
 - ▶ 20 targets have been disambiguated to 2 senses, and 3 targets to 3 senses
 - ▶ Even a **difficult task for humans**, or impossible taken into account in the context words BLESS

dress (clothing)

- **GOLD** *dress*_n¹ *frock*_n² “a one-piece garment for a woman; has skirt and bodice”, *Clothing*_c+
- *attire*_n¹ *garb*_n¹ *dress*_n² “clothing of a distinctive style or for a particular occasion”, *Clothing*_c+
- *apparel*_n¹ *wearing_apparel*_n¹ *dress*_n³ *clothes*_n¹ “clothing in general”, *Clothing*_c+

villa (building)

- *Villa*_n¹ *Pancho_Villa*_n¹ *Francisco_Villa*_n¹ *Doroteo_Arango*_n¹ “Mexican revolutionary leader (1877-1923)”, *Man*_c@
- **GOLD** *villa*_n² “detached or semidetached suburban house”, *Building*_c+
- **GOLD** *villa*_n³ “country house in ancient Rome consisting of residential quarters and farm buildings around a courtyard”, *Building*_c+
- **GOLD** *villa*_n⁴ “pretentious and luxurious country residence with extensive grounds”, *StationaryArtifact*_c+

WSD Results and Comparison

- We have manually constructed a *gold standard* for the 200 targets
- We have disambiguated the 200 targets using UKB (Agirre et al., 2018), which is state-of-the-art in WSD
- We take as baseline a random disambiguator
- Results:

WSD	#	%
Baseline	–	55.38%
UKB	180	90.00%
Adimen-SUMO	142	71.00%

- UKB disambiguates correctly 39 targets that Adimen-SUMO does not
- Adimen-SUMO disambiguates correctly 14 targets that UKB does not
- The upper bound is **97.00%** when using both UKB and Adimen-SUMO

Some Examples

Why are we not able to disambiguate *donkey* (ground_mammal) and UKB is?

- *donkey*_n¹ “the symbol of the Democratic Party; introduced in cartoons by Thomas Nast in 1874”, *Icon*_c+
- **GOLD** *domestic_ass*_n¹ *donkey*_n² *Equus_asinus*_n² “domestic beast of burden descended from the African wild ass; patient but stubborn”, *HoofedMammal*_c+
- **Context** mero: neck, ear, leg... hyper: animal, mammal, vertebrate... event: die, eat, live... coord: fox, lion, pig... and attri: big, large, stubborn...
- **Why???**

Some Examples II

Why are we able to disambiguate *fighter* (vehicle) and UKB is not?

- *combatant_n¹ battler_n¹ belligerent_n¹ fighter_n¹ crapper_n¹* “someone who fights (or is fighting)”, *SocialRole_c+⁺*
- **GOLD** *fighter_n² fighter_aircraft_n¹ attack_aircraft_n¹* “a high-speed military or naval airplane designed to destroy enemy aircraft in the air”, *AirCraft_c+⁺*
- *champion_n² fighter_n³ hero_n³ paladin_n¹* “someone who fights for a cause”, *Human_c+⁺*
- **Context** mero: missile, seat, metal... hyper: plane, vehicle, transport...
event: leave, go, run... coord: car, bus, train... and attri: big, destructive, lethal...
- **Why???**

Some Examples III

Why no approach is able to disambiguate *herring* (water_animal)?

- *herring*_n¹ “valuable flesh of fatty fish from shallow waters of northern Atlantic or Pacific; usually salted or pickled”, Meat_c+
- GOLD *herring*_n² *Clupea_haranguis*_n¹ “commercially important food fish of northern waters of both Atlantic and Pacific”, Fish_c+
- **Context** mero: eye, skin, tail... hyper: **fish, food**, animal... event: **cook, live**, eat (eat or to be eaten?)... coord: salmon, tuna, cod... and attri: edible, fresh, small...
- Both meanings (animal and flesh) are mixed in BLESS dataset

Error Analysis Remarks

- No possible disambiguation if both synsets belong to the same SUMO concept
- Context in BLESS can also be ambiguous (or not so fine grained as in WordNet), e.g. the attri *edible* applied to animals does not help to decide whether it is the animal or its flesh or the trees and fruit or vegetables

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Conclusions

- Adimen-SUMO is subject of improvement
 - ▶ Mapping and knowledge errors
 - ▶ Lack of information (e.g. about events)
- Vampire v4.2.2 is able to work with Adimen-SUMO
 - ▶ Specialized portfolios
 - ▶ Increase resource limits
- Anyway, Adimen-SUMO can be applied to NLP tasks in its current state

Ongoing Work

- We are correcting Adimen-SUMO
 - ▶ Using black-box and white-box techniques (Álvarez et al., 2017a)
 - ▶ Cross-checking Adimen-SUMO with other knowledge resources such as WordNet (Álvarez and Rigau, 2018; Álvarez et al., 2018)
- We are improving the formalization of Adimen-SUMO
 - ▶ By providing an stronger axiomatization: restrictions at the level of classes and not objects
 - ▶ By applying CWA and *Unique Name Assumption* (UNA) (Russell and Norvig, 2010)
- We are trying other ATPs like E (Schulz, 2002), CVC4 (Deters et al., 2014) or iProver (Korovin, 2013)
- We are implementing some optimizations of the resulting FOL formula
 - ▶ Removing non-constant function symbols
 - ▶ Preprocessing

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