

Building and Evaluating a Distributional Memory for Croatian

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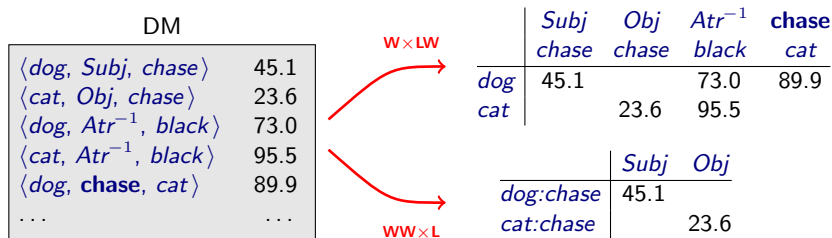
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- Representation of word meaning based on **distributional hypothesis** (Harris, 1954):
 - correlation between similarity of words' contexts and words' semantic similarity
 - words represented as vectors of context features
 - semantic similarity predicted via vector similarity
- **Distributional semantic models** used in many applications (Turney and Pantel, 2010)
- Most models use word-based or syntax-based co-occurrences
- Advantages of syntax-based models:
 - model fine-grained types of semantic similarity
 - capture long-distance contextual relationships
⇒ important for free word order languages
 - applicable to various semantic tasks

Distributional memory (DM) (Baroni and Lenci, 2010)

- General, task-independent framework for distributional semantics
- Set of weighted **Word-Link-Word triplets** obtained from a corpus
 - links can be chosen to model (un)lexicalized dependency relations
- Task-specific sem. spaces obtained by arranging triplets into matrix



- Dependency-based DM for English (Baroni and Lenci, 2010) and German (DM.DE) (Padó and Utt, 2012)

- A challenge, because Croatian is an under-resourced and a morphologically complex language
- Required:
 - good, clean, and large corpus
 - good linguistic preprocessing
- Steps:
 - 1 Corpus preparation
 - 2 Tagging, lemmatization, and parsing
 - 3 Triplet extraction

Step 1: Corpus preparation

- Croatian web corpus **hrWaC** (Ljubešić and Erjavec, 2011)
- Boilerplate removed, but still contains non-parsable content
 - code snippets, encoding errors, non-diacriticized text, foreign-language content (Serbian, Slovenian, English, ...)
- Additional heuristic filtering:
 - 1 website filter: blog/discussion forum content removed
 - 2 document filter: too short, foreign-language
 - 3 sentence filter: too short, non-standard symbols, non-diacriticized, foreign-language
- Filtered corpus **fHrWaC**: 51M sentences and 1.2G tokens

Step 2: Tagging, lemmatization, and parsing

- We trained the models on SETIMES.HR, the Croatian part of the SETimes parallel corpus
 - 90K tokens and 4K sentences
 - manually lemmatized and morphologically annotated
 - dependency annotated by Agić and Merkle (2013)
- HunPos tagger (Halácsy *et al.*, 2007)
- CST lemmatizer (Ingason *et al.*, 2008)
- MSTParser dependency parser (McDonald *et al.*, 2006)

Tagging, lemmatization, and parsing accuracy

		SETIMES.HR	Wikipedia
HunPos (POS only)	Acc	97.1	94.1
CST lemmatizer	Acc	97.7	96.5
MSTParser	LAS	77.5	68.8

- performance on Wikipedia: cross-domain evaluation
- state of the art performance for Croatian
 - see (Agić and Merkle, 2013) and (Agić *et al.*, 2013) for details

Step 3: Triplet extraction

- 10 unlexicalized link types:
 - main dependency relations: *Pred, Atr, Adv, Atv, Obj, Prep, Pnom*
 - subject subcategorization (*Sub_tr/ Subj_intr*) to account for meaning shift due to verb reflexivization
 - predati (to hand in): <student, Subj_tr, predati >*
 - predati se (to surrender): <trupe/troops, Subj_intr, predati >*
 - an underspecified *Verb* link
- 2 lexicalized link types:
 - prepositions: *<mjesto/place, na/on, sunce/sun >*
 - verbs: *<država/state, kupiti/buy, količina/amount >*
- Triplets scored with local mutual information

$$\text{LMI}(w_1, l, w_2) = f(w_1, l, w_2) \log \frac{P(w_1, l, w_2)}{P(w_1)P(l)P(w_2)}$$

Triplet extraction accuracy

Link		P (%)	R (%)	F ₁ (%)
Unlexicalized	<i>Adv</i>	57.3	52.7	54.9
	<i>Atr</i>	85.0	89.3	87.1
	<i>Atv</i>	75.3	70.9	73.1
	<i>Obj</i>	71.4	71.7	71.5
	<i>Pnom</i>	55.7	50.8	53.1
	<i>Pred</i>	81.8	70.6	75.8
	<i>Prep</i>	50.0	28.6	36.4
	<i>Sb_tr</i>	67.8	73.8	70.7
	<i>Sb_intr</i>	64.5	64.8	64.7
	<i>Verb</i>	61.6	73.6	67.1
Lexicalized	Prepositions	67.2	67.9	67.5
	Verbs	61.6	73.6	67.1
All links		73.7	75.5	74.6

- 2.3M lemmas, 121M links and 165K link types
- top-scored (w_1, l, w_2) triplets for $w_1 = \textit{kupiti}$ (to buy) :

l	w_2	LMI
<i>Atv</i>	<i>moći</i> (<i>can_V</i>)	225107
<i>Atv</i>	<i>željeti</i> (<i>wish_V</i>)	22049
<i>Obj⁻¹</i>	<i>stan</i> (<i>apartment_N</i>)	19997
po	<i>cijena</i> (<i>price_N</i>)	18534
<i>Pred</i>	<i>kada</i> (<i>when_R</i>)	14408
<i>Obj⁻¹</i>	<i>dionica</i> (<i>share_N</i>)	13720
<i>Atv</i>	<i>morati</i> (<i>must_V</i>)	12097
<i>Obj⁻¹</i>	<i>ulaznica</i> (<i>ticket_N</i>)	11126
<i>Adv</i>	<i>moguće</i> (<i>possible_R</i>)	9669
<i>Atv</i>	<i>namjeravati</i> (<i>intend_V</i>)	9095
<i>Obj⁻¹</i>	<i>karta</i> (<i>ticket_N</i>)	8936
...

Task-based evaluation

- **Synonym choice** – standard task from distributional semantics

Q: *težak* (*farmer*)

A: (a) *poljoprivrednik* (*agriculturist*) (b) *umjetnost* (*art*)
(c) *radijacija* (*radiation*) (d) *bod* (*point*)

- Dataset: 1,000 question items for nouns, verbs, and adjectives, compiled from a machine readable dictionary (Karan *et al.*, 2012)
- Model: $W \times LW$
- Prediction: Cosine similarity
- Evaluation: Accuracy (%) + Coverage (%)

Synonym choice: Results

Model	Accuracy (%)			Coverage (%)		
	N	A	V	N	A	V
DM.HR	70.0	66.3	63.2	99.9	99.1	100
BOW-LSA	67.2	68.9	61.0	100	100	100
BOW baseline	59.9	65.7	55.9	99.9	99.7	100

- Nearly complete coverage
- Outperforms BOW baseline and performs comparable to LSA
- Differences across POSes
 - nouns: well modeled in syntactic space
 - adjectives: less well modeled (mostly occur with *Atr* links)
 - verbs: poorly modeled in word and syntactic spaces

- DM.HR is a syntax-based DM for Croatian built from a dependency-parsed web corpus
 - first DM for a Slavic language
 - freely available from takelab.fer.hr/dmhr
- Evaluation on synonym choice task
 - DM.HR outperforms BOW, numerically outperforms LSA
- DM.HR can be used for a variety of semantic tasks
- Future work
 - better modeling of adjectives and verbs
 - influence of corpus preprocessing/link types

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