Building and Evaluating a Distributional Memory for Croatian

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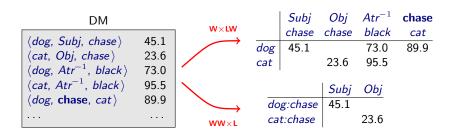
The 51st Annual Meeting of the Association for Computational Linguistics Sofia, August 7, 2013

Distributional semantics

- 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 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)

Building DM.HR

- Required:
 - good, clean, and large corpus
 - good linguistic preprocessing
- A challenge, because Croatian is an under-resourced and a morphologically complex language
- Steps in building DM.HR:
 - Corpus preparation
 - 2 Tagging, lemmatization, and parsing
 - 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:
 - website filter: blog/discussion forum content removed
 - document filter: too short, foreign-language
 - 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

- \bullet We trained the models on ${\rm SETIMES.HR},$ the Croatian part of the SETimes parallel corpus
 - 90K tokens and 4K sentences
 - manually lemmatized and morphologically annotated
 - dependency annotated by Agić and Merkler (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 Merkler, 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): \(\strupe/troops, Subj_intr, predati \)
 - an underspecified Verb link
- 2 lexicalized link types:
 - prepositions: \(\text{mjesto/place}, \, \text{na/on}, \, \text{sunce/sun} \)
 - verbs: \(\langle država \state, \) kupiti/buy, \(količina \slangle amount \) \)
- Triplets scored with local mutual information

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

Triplet extraction accuracy

Link		P(%)	R(%)	F ₁ (%)
Unlexicalized	Adv	57.3	52.7	54.9
	Atr	85.0	89.3	87.1
	Atv	75.3	70.9	73.1
	Obj	71.4	71.7	71.5
	Pnom	55.7	50.8	53.1
	Pred	81.8	70.6	75.8
	Prep	50.0	28.6	36.4
	$Sb_{-}tr$	67.8	73.8	70.7
	Sb_intr	64.5	64.8	64.7
	Verb	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

DM.HR

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

l	w_2	LMI
Atv	moći (can $_{V}$)	225107
Atv	željeti $({\it wish}_V)$	22049
Obj^{-1}	$stan\ (apartment_N)$	19997
ро	$\mathit{cijena}\;(\mathit{price}_N)$	18534
Pred	$\mathit{kada}\;(\mathit{when}_R)$	14408
Obj^{-1}	$dionica$ $(share_N)$	13720
Atv	morati (must $_{V}$)	12097
Obj^{-1}	ulaznica (ticket $_N$)	11126
Adv	${\sf mogu\'ee}$ (possible $_R$)	9669
Atv	namjeravati (intend $_{V}$)	9095
Obj^{-1}	$\mathit{karta}\ (\mathit{ticket}_N)$	8936

Task-based evaluation

Synonym choice – standard task from distributional semantics

Q: težak (farmer)

- (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×IW
 - Prediction: Cosine similarity
 - Evaluation: Accuracy (%) + Coverage (%)

Synonym choice: Results

	Accuracy (%)			Coverage (%)		
Model	N	А	V	N	А	V
Dm.HR	70.0	66.3	63.2	99.9	99.1	100
LSA (Karan <i>et al.</i> , 2012)	67.2	68.9	61.0	100	100	100
BOW baseline	59.9	65.7	55.9	99.9	99.7	100

- Outperforms BOW and numerically outperforms LSA on N and V
- 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
- Nearly complete coverage

Summary

- 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
 - $\bullet~{\rm DM.HR}$ outperforms BOW, numerically outperforms LSA on N and V
- 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

Acknowledgment



This work was supported by the Croatian Science Foundation under the grant 02.03/162: "Derivational Semantic Models for Information Retrieval"

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