



Detection of Non-recorded Word Senses

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Intro

- Unknown Sense Detection (USD)
- Support dictionary maintenance
- Realistic few-shot scenario
- English and Swedish
- Modern and historical



Task

Data

Models

Experiments

Conclusion

References

Unknown Sense Detection (USD)

USD is the identification of corpus occurrences that are not covered by a given sense inventory. (Erk (2006))

- Related to other popular tasks like Word Sense Detection (WSD)
- A specific task of Natural Language Processing (NLP)
- An underlying sense inventory
- Word usages to examine
- A methodology to recognize and differentiate between the meanings of words (in context)

Lexical Resources (WordNet)

- A well-known and well-established large lexical database for the English language.
- It groups words which are synonymous in one of their meanings into so-called synsets that describe distinct concepts or senses. <car, auto, automobile, machine, motorcar> <cable car, car>
- ► A total of 117,000 synsets, all containing a gloss.
- Only a portion includes example usages

Lexical Resources (Svensk Ordbok (SO))

- A dictionary of the Swedish language, created and maintained by the University of Gothenburg (Allén, 1981).
- SO is privately managed
- Lists senses under a headword
- A total of 68,000 senses for over 41,500 headwords
- The majority of the senses are described by a gloss
- All sense entries contain example usages

Corpora (Modern)

English		Swedish		
Name	Leipzig_News	Leipzig_News		
Year	2020	2022		
Source	Goldhahn, Eckart	, and Quasthoff (2012)		
Sentences	1 million	1 million		

Table 1: Comparison of modern corpora

Corpora (Historical)

	English	Swedish	
Name	ССОНА	Kubhist2	
Year	1810–1860	1790–1830	
Source	Alatrash et al.	Språkbanken	
	(2020)	(downloaded in	
		2019)	
Sentences	250 thousand	3.3 million	

Table 2: Comparison of historical corpora

Contextualized Embeddings (XL-LEXEME)

- SentenceBERT (SBERT): extends BERT, using siamese and triplet networks to produce meaningful sense embeddings for a single sentence (Reimers & Gurevych, 2019)
- XL-LEXEME is a pretrained contextualized embeddings model that is based on SBERT and fine-tuned on human-labeled Word-in-Context (WiC) data (Cassotti et al., 2023)
- Gives prominence to a target word in the given context.

Models (Sense inventory)

- Solely data from the dictionary
- Gloss and example word usages undergoing different replacement strategies (Table 3)
- Naturally limited in their predictions by gaps in the data, i.e., missing gloss or example usages

option	description	example
0	leaves the sequence as is	a poor salary
1	HEADWORD: SEQUENCE	inadequate: a poor salary
2	SEQUENCE (HEADWORD)	a poor salary (inadequate)
3	SEQUENCE, i.e., HEADWORD	a poor salary, i.e., inadequate
4	replace word	a inadequate salary

 Table 3: Replacement strategies on an example of *inadequate* taken from

 WordNet

Models (Target embeddings and comparision)

- Word usages from corpora
- Original sentence
- Target token substituted by its lemma
- Contextualized embedding using XL-LEXEME

- Calculate distance to all eligible sense embeddings
 - Cosine similarity
 - Spearman's rank correlation coefficient
- Assign based on a threshold

Experiments (Outline)

- Human annotation for model tuning and to establish a reference point
- Tuning of hyperparameters, i.e., the similarity threshold
- Predicting with the most promising models
- Determine the models' performance with human annotation
- Outline human annotations:
 - A total of six annotators, three for each language
 - All participants are students and native speakers of the respective language
 - They received a 30-minute briefing
 - They underwent a short test annotation

Experiments (Outline human annotation)

- Carried out using the PhiTag platform
- Inspired by Erk et al. (2013)'s WSsim method
- Individual assessment of all senses for a word usage
- Ask for binary classification
 - sense gloss fits (label "1")
 - sense gloss does not fit (label "0")
 - no specification is possible (label "-")

Experiments (PhiTag Interface)

usage:

Abigail's sister accompanied her to this meeting because the sister alleged that she had witnessed some of the childhood sexual abuse from the **relative**.

gloss:

estimated by comparison; not absolute or complete

all glosses:

estimated by comparison; not absolute or complete

an animal or plant that bears a relationship to another (as related by common descent or by membership in the same genus)

a person related by blood or marriage

Figure 1: Illustration of the PhiTag user interface

Experiments (Human annotation: Part 1)

- Human annotation is conducted on a random sample from the corpora
- Identically for both languages
- Word usages were retrieved by lemmatizing the sentences and searching for appearances of a randomly selected headwords
- At most five usages for each headword, chosen at random
- Approximately 1,200 annotation instances, distributed over 500–700 usages
- Only primary synsets are considered

Experiments (Human annotation: Part 1)

- Aggregate judgements per annotation instance by majority
- A usage is considered assigned iff at least one instance has the majority label "1"

	English	Swedish
Instances	1165	1202
Usages	474	706
Label distribution (0, 1, -)	(1840, 1651, 4)	(1294, 2104, 208)
Excluded instances	2	87
Remaining usages	473	674
Assigned	428	562
Unassigned	45 (9.5%)	95 (16.6%)

Table 4: Statistics of the first human annotation phase

Experiments (Threshold tuning)

- ▶ 10 rounds of 5-fold cross-validation for every model
- Same simulated data for all models
- ▶ Try every threshold $\in \{0.0, 0.01, 0.02, ..., 1.0\}$
- Choose threshold to maximize F_{β} -score (including $\beta = 0.3$)

Experiments (Threshold tuning)

Round 10 - E4 - Cosine Similarity



Figure 2: Precisions and recalls of all five folds in the cross-validation round 10 of model E4_COS on English data

Experiments (Threshold tuning)

F _{0.3} -score									
	English					Swedish			
	DEFAULT SUBST.		DEFA	DEFAULT		SUBST.			
	COS	SPR	COS	SPR	COS	SPR	COS	SPR	
E0	0.466	0.523	0.477	0.452	0.417	0.419	0.413	0.400	
E 1	0.589	0.590	0.583	0.592	0.451	0.443	0.413	0.399	
E2	0.579	0.562	0.566	0.590	0.425	0.431	0.411	0.408	
E3	0.494	0.523	0.493	0.489	0.428	0.431	0.397	0.392	
E4	0.613	0.584	0.593	0.612					
G0	0.270	0.280	0.255	0.267	0.349	0.371	0.345	0.340	
G1	0.227	0.220	0.226	0.209	0.600	0.606	0.549	0.537	
G2	0.260	0.223	0.264	0.245	0.550	0.612	0.564	0.547	
G3	0.217	0.259	0.275	0.256	0.625	0.617	0.599	0.621	

Table 5: Results of cross-validation. Performance is given as average $F_{0.3}$ across rounds and folds

Experiments (Human annotation: Part 2)

- Prediction on 100–150 thousand roughly filtered sentences, from both modern and historical corpora, respectively.
- Every sentence is lemmatized and then searched for headwords represented in the respective model's sense inventory.
- A prediction is made for all usages of the found headwords
- Exclude partially complete headwords (at least one sense incomplete)
- Unassigned usages are sorted by similarity to the nearest sense
- At most eight usages for the same headword are chosen from the top of this list

Experiments (Human annotation: Part 2)

- Aggregate judgements per annotation instance by majority
- A usage is considered assigned iff at least one instance has the majority label "1"

	English	Swedish
Instances	1208	1400
Usages	322	1001
Label distribution (0, 1, -)	(2151, 1462, 11)	(2529, 1218, 456)
Excluded instances	5	109
Remaining usages	322	927
Assigned	277	327
Unassigned	45 (13.98%)	600 (64.725%)

Table 6: Statistics of the second human annotation phase

Evaluation (Comparison)

		Phase 1		Evaluation Phase		
Usages	All	Modern	Historical	All	Modern	Historical
total	474	232	242	322	210	112
excluded	1	1	0	0	0	0
remaining	473	231	242	322	210	112
assigned	428	208	220	277	176	101
unassigned	45 (9.5%)	23 (9.9%)	22 (9.1%)	45 (13.98%)	34 (16.2%)	11 (9.8%)

	Phase 1				Evaluation Phase			
Usages	All	Modern	Historical	All	Modern	Historical		
total	706	337	369	1001	478	523		
excluded	52	4	28	74	9	65		
remaining	674	333	341	927	469	458		
assigned	562	293	269	327	224	103		
unassigned	112 (16.6%)	40 (12.0%)	72 (21.1%)	600 (64.7%)	245 (52.2%)	355 (77.5%)		

Table 7: Comparison of the different corpora of the English and Swedish data

Cases where also a close manual analysis suggests that they are truly non-recorded in our dictionary:

usage

No wonder hes up there getting big **baked**.

senses

- cook and make edible by putting in a hot oven
- prepare with dry heat in an oven
- heat by a natural force
- be very hot, due to hot weather or exposure to the sun
- dried out by heat or excessive exposure to sunlight
- (bread and pastries) cooked by dry heat (as in an oven)

Cases with likely historical meanings that are seemingly non-recorded:

usage

You seem to intend a eulogy, yet leave out whatever was noblest in her, and blacken while you mean to praise.

senses

make or become black

burn slightly and superficially so as to affect color

A major problem are multi-word expressions:

usage

Im at that age where many of my friends are having children, and a central topic of conversation whenever were together **revolves** around creating the almostscientifically set schedule for their babies.

senses

turn on or around an axis or a center

move in an orbit

cause to move by turning over or in a circular manner of as if on an axis

Occurrences of ellipsis of a represented multi-word:

usage

Notably, the local Native American tribes were not targeted or **wiped** the new nation embraced them as equals and allowed the tribes a major say in the rule of America.

senses

rub with a circular motion

OCR and spelling errors:

 $``[\dots]$ mud turtles , and floating timber to say nothing of water snakes , which were far more terrible to me than shirks .''

Likely ambiguous usages:

"Stolen bases are a thing for me, Betts said."

Incorrect lemmatization:

virtually → virtual

Ad-hoc innovative word meanings:

"But sloppy, agendadriven journalism of this sort fans the **flames** of racial division."

Conclusion

- Automatically detect non-recorded word senses in a large corpora, based on a given large, but possibly imperfect sense inventory.
- The method considerably increases the chance to find non-recorded word senses in corpus usages compared to a random baseline
- Predict a large number of unassigned usages that can be used to update WordNet's and SO's sense inventory in the near future
- Manual analysis shows some weaknesses like faulty multi-word detection
- No manual analysis of the Swedish data yet

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