

RuSemCor: A Word Sense Disambiguation Corpus for Russian HSE University

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Summary

We present RuSemCor, an open Word Sense Disambiguation (WSD) corpus for Russian. The corpus was constructed by manually linking tokens from the OpenCorpora corpus to senses in the Russian wordnet RuWordNet. It consists of 869 documents with 121,710 tokens of which 51,588 are wordnet annotated. The resource is represented using the NIF, OLiA, OntoLex, and Global WordNet ontologies and integrated into the Linguistic Linked Open Data cloud.



We used RuSemCor as a diagnostic benchmark to evaluate a range of WSD methods. Our experiments yielded three main findings. 1) Generative LLMs substantially outperform traditional knowledge-based methods such as Personalized PageRank. 2) Despite their strengths, generative LLMs do not surpass encoder-based models specifically trained for WSD. 3) Incorporating lexical-semantic relations from RuWordNet produces mixed results: it enhances the performance of encoder-based models and leading LLMs like GPT- 4, DeepSeek, and Mistral 24B, but tends to degrade accuracy for smaller generative models such as GPT-3 and Mistral 7B.

The resource is distributed under the CC BY-SA open license and is available at:

http://github.com/llod-ru/rusemcor

RuSemCor statistics

| Entity | Count |
|-----------------------------|---------|
| Documents | 869 |
| Sentences | 7,398 |
| Tokens | |
| All tokens | 121,710 |
| Sense-annotated tokens | 51,588 |
| Lemmas | |
| All lemmas | 18,244 |
| Sense-annotated lemmas | 10,680 |
| Covered RWN synsets | 9,075 |
| Covered RWN lexical entries | 9,211 |
| Covered RWN senses | 11,152 |
| | |

Corpus construction

- 1. Automatic Task Generation: For each token, a list of candidate RWN synsets was generated by aligning via lemma and POS. This required resolving differences in POS inventories and lemmatization between the two resources.
- **2. Sampling**: 869 mid-sized documents were manually sampled for annotation.
- 3. Manual Annotation: Three trained annotators, supervised by an expert, selected the correct synset for each token. High inter-annotator agreement (96% F1) was achieved. Note: Not all tokens are linked to RWN, due to: missing POS categories, missing lexical entries, or no suitable synset in context.
- **4. Final Assembly**: The completed annotations were processed into the final corpus, combining sense annotations from RWN with text metadata and morphology from OpenCorpora.

Corpus description

RuSemCor is a corpus of Russian texts with morphological and lexical-semantic annotations, built from OpenCorpora by manually linking its tokens to senses in RuWordNet.

Source Resources:

- **OpenCorpora**: An open Russian corpus with morphological annotations (lemmas, POS tags, grammatical features).
- RuWordNet (RWN): A Russian wordnet with synsets for nouns, verbs, and adjectives.

Experimental setups (WSD)

To investigate whether other structured information from the wordnet graph — especially semantic relations — can improve disambiguation accuracy, we designed and evaluated several model setups. Each setup constructs the sense description from different combinations of synset features derived from RuWordNet. **These features are**: artificial numerical ID (id), canonical name (name), definition (def), lexical entries (entries), hyponyms (hypos), hypernyms (hypers), and the hyponym—hypernym chain from the synset to the root of the graph, represented both as se- quences of synset names (chain) and of numerical IDs (idchain).

For this evaluation, we adapted **ConSeC** to Russian by employing the pre-trained model deepvk/deberta-v1-base as the encoder backbone. This model is a Russian-language DeBERTa-v1-base variant pre-trained on 400 GB of diverse Russian texts, including Wikipedia, books, social media, and news. With this backbone we fine-tuned ConSeC on the training subset of our RuSemCor corpus.

Prompt for LLMs

"Consider the following sentence: «Право на жизни находится под угрозой во всем мире, говорит Папа.». In what sense the word «Папа» is used in this sentence:

- 1) The sense «ОТЕЦ», that can be expressed by the following terms: БАТЯ, ПАПА, ПАПУЛЯ, БАТЮШКА, ПАПОЧКА, ОТЕЦ, ПАПАША, БАТЬКА, ПАПЕНЬКА. The hypernyms of this sense are the following: РОДСТВЕННИК-МУЖЧИНА, РОДИТЕЛЬ. The hyponyms of this sense are the following: ПРАДЕД, ОТЕЦ МУЖА ИЛИ ЖЕНЫ.
- 2) The sense «ПАПА РИМСКИЙ», that can be expressed by the following terms: ПАПА РИМСКИЙ, ПАПА, ПОНТИФИК, РИМСКИЙ ПАПА, СВЯТЕЙШЕСТВО, СВЯТОЙ ПАПА, ПАПСТВО, ЕГО СВЯТЕЙШЕСТВО, СВЯТЕЙШИЙ ПАПА, ГЛАВА КАТОЛИЧЕСКОЙ ЦЕРКВИ. The hypernyms of this sense are the following: ГЛАВА ГОСУДАРСТВА, ГЛАВА ЦЕРКВИ, КАТОЛИЧЕСКИЙ СВЯЩЕННИК. The definition of this sense is «верховный глава католической церкви; папа». Please, print only number of the appropriate sense and nothing else."

Results of evaluation of WSD methods

| Method | Accuracy (%) |
|-------------------------------------|--------------|
| Baseline | , , |
| Random | 63.60 |
| Min id | 70.91 |
| Most frequent | 74.10 |
| Most frequent + min id | 84.54 |
| | |
| Knowledge-based | |
| Personalized PageRank | 74.30 |
| | |
| Finetuned encoder | |
| ConSeC | 07.00 |
| id | 87.88 |
| name | 91.25 |
| idchain | 92.57 |
| | 92.20 |
| name+def | 92.42 |
| name+def+entries+hypos+hypers | 92.70 |
| Generative LLM | |
| GPT-3.5 | |
| name+def | 83.95 |
| name+def+entries+hypos+hypers | 82.92 |
| Mistral 7B Instruct v0.3 | 02.32 |
| name+def | 83.77 |
| name+def+entries+hypos+hypers | 83.18 |
| name+def+entries+hypos+hypers+chain | 82.06 |
| Mistral Small 24B Instruct 2501 | |
| name | 89.52 |
| def | 78.42 |
| entries | 85.32 |
| hypos | 80.77 |
| hypers | 83.07 |
| hypos+hypers | 86.57 |
| chain | 88.12 |
| name+def | 89.21 |
| name+def+entries | 90.15 |
| name+def+hypos+hypers | 90.88 |
| name+def+chain | 90.05 |
| name+def+entries+hypos+hypers | 90.80 |
| name+def+entries+hypos+hypers+chain | 90.63 |
| Deepseek | |
| name | 91.50 |
| def | 76.86 |
| entries | 86.50 |
| hypos | 81.86 |
| hypers | 82.75 |
| hypos+hypers | 87.87 |
| chain | 90.21 |
| name+def | 91.15 |
| name+def+entries | 91.41 |
| name+def+hypos+hypers | 91.99 |
| name+def+chain | 91.43 |
| name+def+entries+hypos+hypers | 91.96 |
| name+def+entries+hypos+hypers+chain | 91.88 |
| GPT-4 | |
| name+def | 91.56 |
| name+def+entries+hypos+hypers | 92.09 |

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