

On Ellipsis in Slavic: The Ellipsis Corpus and Natural Language Processing Results

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Team

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Slides and publications, and links to more material available at:

https://nlp-lab.org/



Agenda

- Ellipsis Constructions and Syntax
- The Hoosier Ellipsis Corpus
- Slavic Sub-Corpora
- Machine Learning Experiments



SECTION 1

Introduction and Motivation

- Common phenomena like gapping, sluicing, forward or backward conjunction reduction
 - Lexical elements are elided under certain conditions (e.g., syntactic, discourse)
 - Native speakers have no cognitive issues processing and understanding ellipsis constructions
 - Ellipsis constructions are very frequent in common genres

Examples...



Sluicing:

- Moja sestra živi u Utrechtu ali ne znam gdje ____
 - → Moja sestra živi u Utrechtu ali ne znam gdje (moja sestra u Utrechtu živi)

Gapping

- Иван и Андрей смотрели новости, а Ольга ___ фильм.
 - → Иван и Андрей смотрели новости, а Ольга <mark>смотрела</mark> фильм.
- Czy Marek zobaczył Annę pierwszy, czy Anna ____ Marka ____ ?
 - → Czy Marek zobaczył Annę pierwszy, czy Anna zobaczyła Marka pierwsza ?



- Discourse Licensed Ellipsis:
- A: Tko želi sresti koga?
- B: Suzana ____ Petra.
 - → Suzana <u>želi sresti</u> Petra.
- Semantic Issues:
- Marek pojechał do Warszawy i ____ został aresztowany w Poznaniu.
- Peter stole a book and John ____ kisses from Mary. (zeugma, Sennet 2016)



- Publicly available datasets:
 - Sluicing corpus for English
 - VP-ellipsis corpus for English
 - ELLies corpus for English
- Small datasets
- Limited to English and a few common languages
- Limited to specific ellipsis phenomena (gapping, sluicing, or VP-ellipsis)



- Lack of a cross-linguistic typological overview of ellipsis types
- Explanatory theoretical analysis of ellipsis constructions
- Frameworks like Dependency Grammar or Lexical-functional Grammar do not provide descriptive or explanatory means
 - even Generative frameworks like Minimalist Program do not general explanations

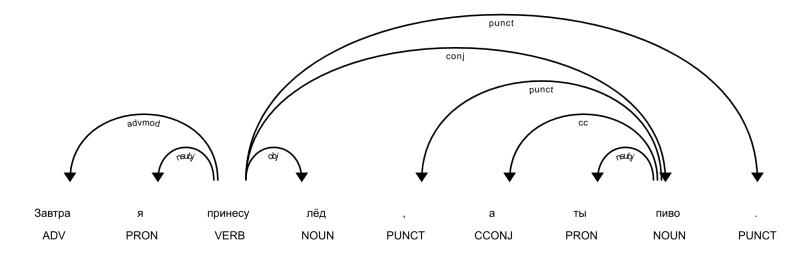


Ellipsis Constructions: Classic NLP

- Current State of the Art (SOTA) Natural Language Processing-pipelines and parsers perform poorly (or not at all)
- Tested SOTA parsers:
 - Stanford CoreNLP
 - Stanford Stanza (V 1.8.2) (Dependency & Constituent Parser)
 - Berkley Neural Parser (benepar) (V 0.2.0)
 - SpaCy 3.7
 - XLE (Web-XLE, Lexical-functional Grammar Parser)
- All parsers fail with Ellipsis (and other constructions) → not useful for downstream NLP tasks (e.g., relation extraction)



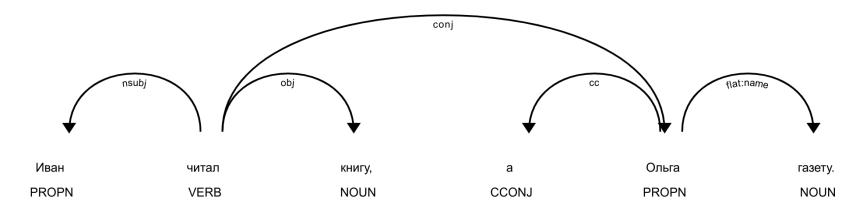
Dependency Parsers



Stanza 1.8.2
Resulting assumption:
I will bring ice and beer (you?); coordination of "bring" and "beer"



Dependency Parsers

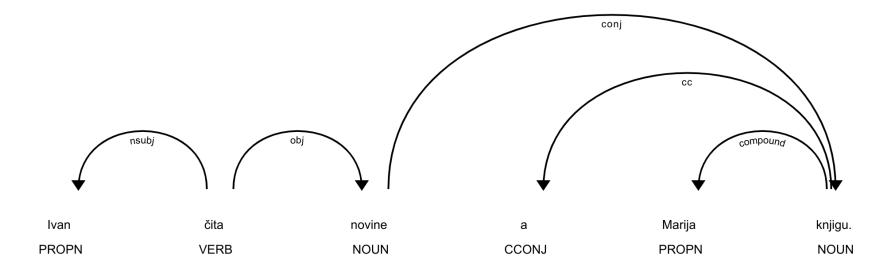


Stanza 1.8.2
Resulting assumption:

Coordination of "read" and "Olga" - compound direct object newspaper



Dependency Parsers



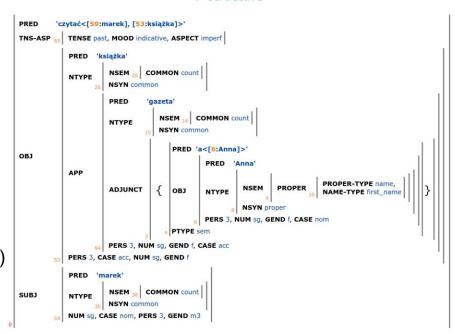
Spacy 3.7
Result: Coordination of "newspaper" and "book"



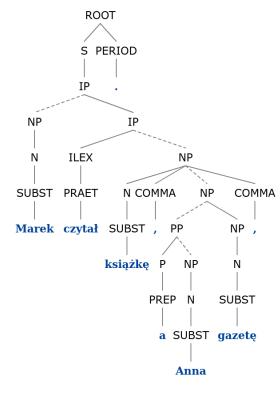
C-structure

LFG Parser

F-structure

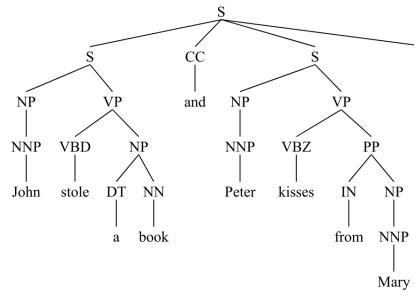


XLE Web Polish (POLFIE)





Constituent Parsers



Berkley Neural Parser Head Noun of the object (kisses) is assumed to be the predicate head of the second conjunct.



Computational Methods and Experiments

Cloze test:

- 1. Used in Machine Learning Masked Word Prediction in BERT (LM)
 - The house ____ I was born. (a. where , b. which)

LLMs:

1. Next word prediction as in GPT and other Large Language Models (LLMs)

Tasks:

- 1. Classification of sentences / utterances: Does it contain ellipsis or not?
- 2. Detection of locus of ellipsis: indicate the space
- 3. Guess of the missing words: fill in the missing words



- 18 Languages with varying numbers of examples.
 - O Largest: Russian, Polish, Ukrainian
 - O Included: Croatian, Navajo, Gujarati, Chinese, Arabic, English, Spanish...
- Picked:
 - Up to 500 target sentences
 - O 1000 or more distractors
 - O For task 2 & 3: only examples with ellipsis are used.
- Algorithms:
 - O Logistic Regression
 - O BERT/RoBERTa-based Deep Learning model
 - O GPT-4 Large Language Model (ChatGPT), Claude 3, Falcon2, Llama2, etc.
 - Using API = no memory related to testing N examples



SECTION 2

Data Collection

Building the Corpus

Sources:

- Literature (peer-reviewed articles and scientific books, literature)
- Corpora (valid and checked)
 - Croatian Language Corpus (Literature)
 - National Corpus of Polish

Polish 139 Russian 202 Ukrainian 158

Arabic (375), English (267), German (79), Gujarati (9), Hindi (127), Japanese (105), Korean (40), Kumaoni (85), Mandarin Chinese (40), Navajo (9), Norwegian (55), Spanish (171), Swedish (20), Telugu (20), Croatian, Bosnian, Serbian, Slovenian, Slovak...



Data Structure

```
A Nina ___ na pianinie.

----

A Nina gra na pianinie.

B: Kasia gra na klarnecie.

A: Marek śpiewa.

# source: Marjorie J. McShan (2000)

# TR eng: Nina plays piano.
```

Polish THEC gapping example with additional information.



Data Structure

The THEC data format allows for:

- adding syntactic tree annotations (bracketed notation, triples for dependencies, c- and f-structure)
- Easy processing with our own Python tools, simple scripts
- Extensible to allow for other linguistic annotations
- Simplifies evaluation using classical NLP-technologies and Als/Large Language Models



Corpus Access

- In the next days: See NLP-Lab page
 - https://nlp-lab.org/ellipsis/
- Link to GitHub, allowing for collaboration and contribution.
 - https://github.com/dcavar/hoosierellipsiscorpus
 - Contribution and feedback welcome!



SECTION3

Experiments

Goals

NLP Problems:

- Failing parsers and lack of downstream processing → semantic and pragmatic analysis
- Sentences with ellipsis undone parse mostly correctly
- → Undo ellipsis
- Assumption: ellipsis constructions do differ from unelided constructions, but syntactically this
 makes no difference, if ellipsis information is not lost and feeds the semantic mapping process

Example of ellipsis affecting quantifier scope or interpretation:

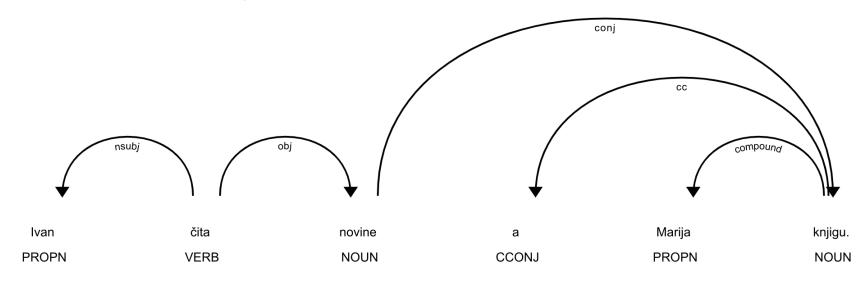
Nobody was smoking and ____ drinking.

Nobody was smoking and nobody was drinking.



Example

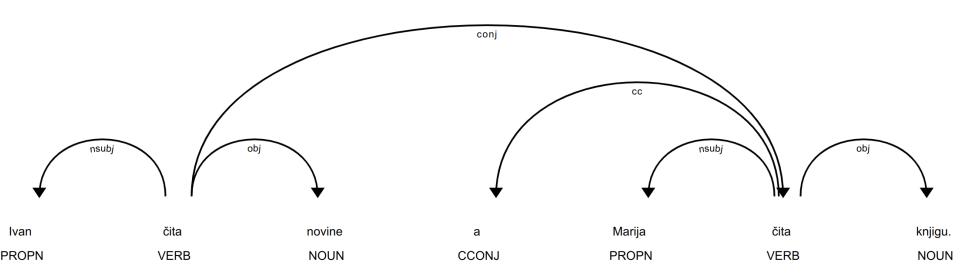
Croatian: spaCy 3.7 Large Croatian Model





Example

Croatian: spaCy 3.7 Large Croatian Model





Task 1: Does the sentence contain ellipsis?

Task 2: Where is the ellipsis? (growing complexity: one, two, three slots)

Task 3: What are the missing words / phrases?

Baseline: Logistic Regression using 8 basic features

- the number of nouns/NPs, subject dependency labels, object dependency labels, conjunctions, verbs, auxiliaries
- a boolean whether an interrogative pronoun is sentence/clause-final position



English in comparison:

- Task 1:
 - Logistic Regression (baseline): accuracy 72%
 - BERT-based Transformer: accuracy 94%
 - GPT-3.5: accuracy: 35%
 - GPT-4: accuracy: 60%

BERT/Transformer > Logistic Regression > GPT-4



- For Arabic:
 - We utilized GPT-4 (no other LLM was capable of processing Arabic)
 - Missing useful BERT-type LM for Arabic, we need to train one
 - Task 1: 0-shot classification
 - Baseline: Logistic Regression 83%
 - GPT-4: Precision 0.56, Recall 0.18
 - Task 3: 0-shot word filling (single word task)
 - Accuracy ~80%



	Russian	Polish	Ukrainian
LogReg	0.63	0.76	0.76
Task 1			
GPT-4 0-shot	0.74	0.73	0.7
GPT-4 few-shot	0.77	0.84	0.8
Task 2			
GPT-4 0-shot	0.28	0.44	0.35
GPT-4 few-shot	0.28	0.36	0.5
Task 3			
GPT-4 0-shot	0.2	0.26	0.24
GPT-4 few-shot	0.32	0.29	0.4



Conclusion

- Problems with "invisible words" in all parsers and LLMs
 - Parsers perform without a problem with "ellipsis undone"
- The problem is:
 - Theoretical Dependency Grammar, Lexical-functional Grammar, etc.
 - Data-based missing corpora with annotated ellipsis constructions
 - Computational LLMs predict next words, and not next missing words (while BERT is trained on masked words)



Conclusion

Goal:

- We are developing tools to detect and undo ellipsis and to use classical NLP for representation learning/generation = syntactic trees, to build semantic representations in a subsequent step
- Work on alternative computational grammar formalisms to handle ellipsis (as well as other construction types: islands, LDDs, discontinuities)





Thank you for your attention!