

Computational Approaches for Understanding Semantic Constraints on Two-termed Coordination Structures

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Part 1: Introduction

- Two-termed
Coordination
- LCL
- Motivation and
Goal

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→ Two-termed Coordination

- LCL
- Motivation and Goal

Coordination groups two or more phrases together:

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Coordination groups two or more phrases together:

Would you like
[_{NP} soup] or [_{NP} salad]?

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Conjuncts

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Coordination groups two or more phrases together:

Would you like

[_{NP} soup] or [_{NP} salad]?

Conjuncts

Coordinating
Conjunction

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Law of Coordination of Likes (LCL)
mandates **like** conjunct categories

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Law of Coordination of Likes (LCL)

mandates **like** conjunct categories

What about ungrammatical **like** coordinations...

* John ate with [_{NP} his mother] and [_{NP} good appetite].

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Motivation: computational approaches have revealed syntactic patterns in coordination

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Goal: what can computational approaches tell us about semantics?

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Motivation: computational approaches have revealed syntactic patterns in coordination

Goal: what can computational approaches tell us about semantics? Patterns in **conjunction** or **categories**?

Part 2: Background and Related Work

- Basic Types of Coordination
- Computational Methods

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Basic types of coordination,
distinguished by conjunction:

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- *and*: links equal elements

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 - *or*: mutually-exclusive options

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Basic types of coordination,
distinguished by conjunction:

- *and*: links equal elements
- *or*: mutually-exclusive options
- *but*: displays semantic contrast

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And-coordinations have the most general meaning and can take different connotations in context

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And-coordinations have the most general meaning and can take different connotations in context

Contrastive temporal sequence:

She [_{VP} tried hard] **but** [_{NP} failed].

Part 2: Background and Related Work

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And-coordinations have the most general meaning and can take different connotations in context

Contrastive temporal sequence:

She [_{VP} tried hard] **and** [_{NP} failed].

Part 2: Background and Related Work

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Or-coordinations are not always disjunctive

Not mutually-exclusive options:

Do you have any

[_{NP} brothers] **or** [_{NP} sisters]?

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Our previous computational approaches focused on **syntax**

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Our previous computational approaches focused on **syntax**

LCL does not account for grammatical, syntactically **unlike** coordinations...

John is [_{AP} healthy] and [_{PP} in good shape].

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Our previous findings show anti-symmetry of coordination syntax:

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- **NPs** tend to coordinate with **subordinate clauses**

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Our previous findings show anti-symmetry of coordination syntax:

- **NPs** tend to coordinate with **subordinate clauses**

I agree with [_{NP} the president]
and [_{SBAR} what he did].

Part 2: Background and Related Work

- Basic Types of Coordination
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Our previous findings show anti-symmetry of coordination syntax:

- The 1st conjunct tends to be **shorter** than the 2nd conjunct
- Support for theories of grammatical weight

Part 3: Approach

- Overview
- UD corpora
- Coordination
Extraction
- Semantic
Analysis

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Key Idea: process and analyze corpus data to study the *semantic* properties of coordination in English

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Key Idea: process and analyze corpus data to study the *semantic* properties of coordination in English

- Extracting coordinations from **Universal Dependencies** corpora

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Key Idea: process and analyze corpus data to study the *semantic* properties of coordination in English

- Extracting coordinations from **Universal Dependencies** corpora
- Measure relatedness using **WordNet** and **word embeddings**

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Universal Dependencies (UD):

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Universal Dependencies (UD):

- Provides consistent dependency annotation across sources

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Universal Dependencies (UD):

- Provides consistent dependency annotation across sources
- Enhanced dependencies of UD v2 augment the conjunct relations

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UD data sources:

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UD data sources:

- English Web Treebank (EWT)

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UD data sources:

- English Web Treebank (EWT)
- Georgetown University Multilayer
Corpus (GUM)

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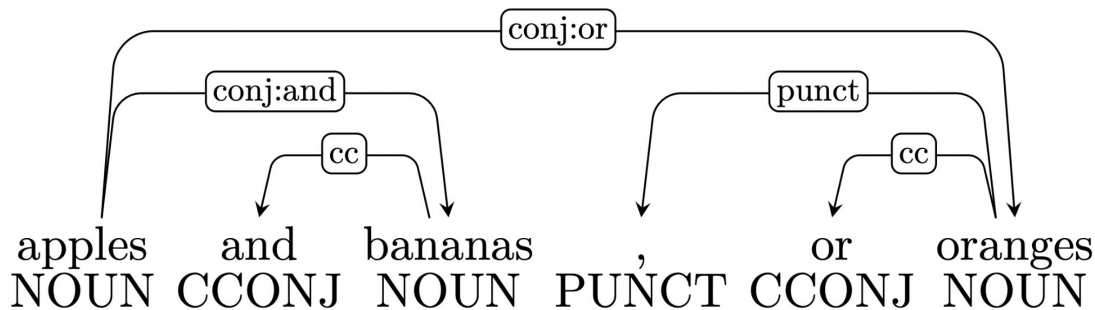
UD data sources:

- English Web Treebank (EWT)
- Georgetown University Multilayer
Corpus (GUM)
- English Parallel Universal
Dependencies (PWT) treebank

Part 3: Approach

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- **Coordination Extraction**
- Semantic Analysis

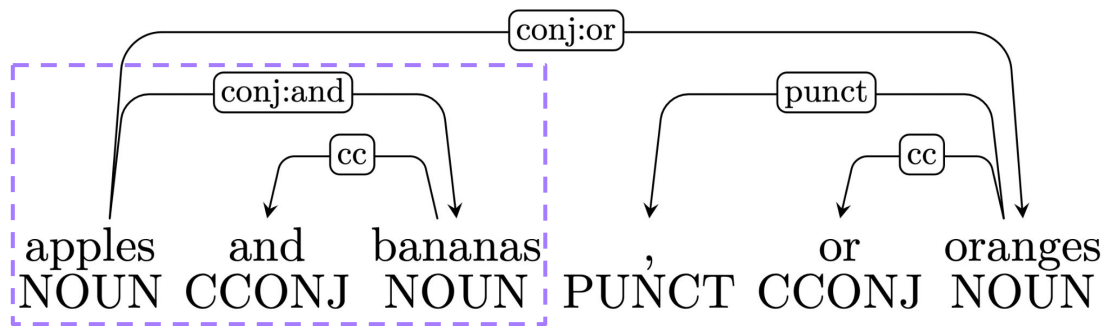
Coordination Extraction:



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Coordination Extraction:

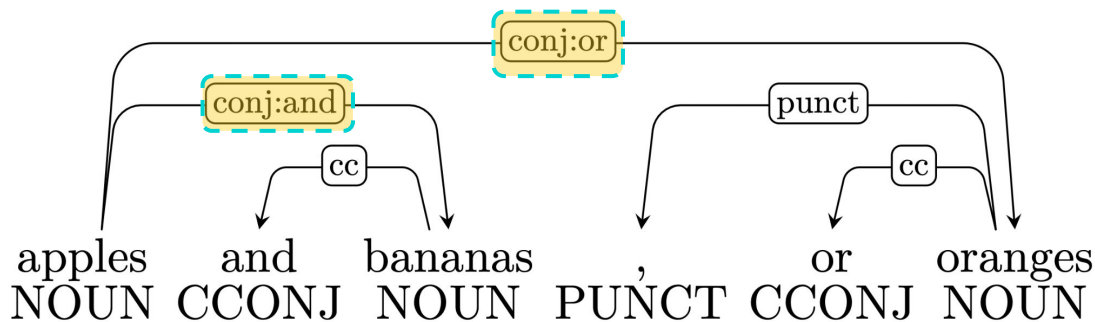


Nested Phrase

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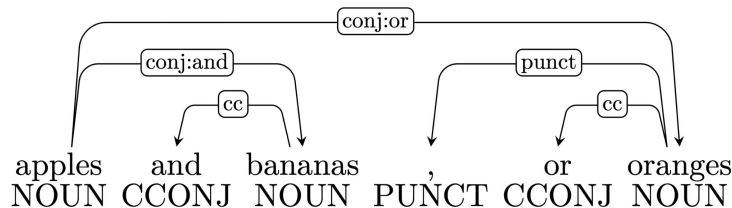
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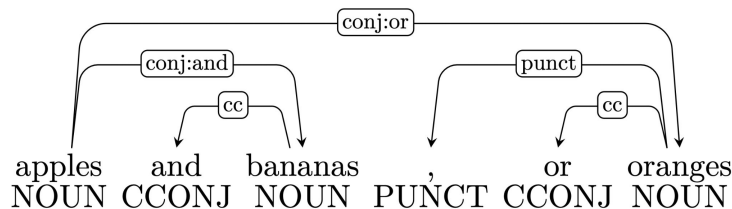
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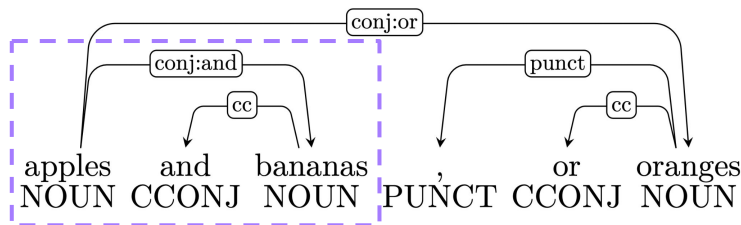


| 1st Conjunct Category | 1st Conjunct Text | 2nd Conjunct Category | 2nd Conjunct Text | Conjunction |
|-----------------------|-------------------|-----------------------|-------------------|-------------|
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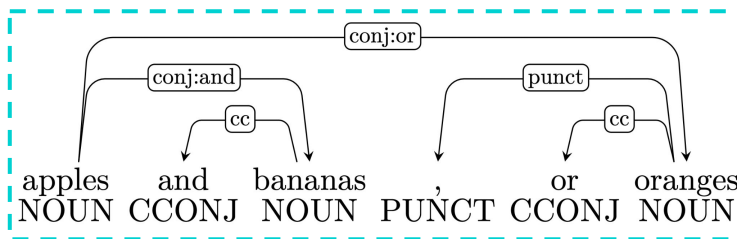


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|-----------------------|-------------------|-----------------------|-------------------|-------------|
| NOUN | apples | NOUN | bananas | and |

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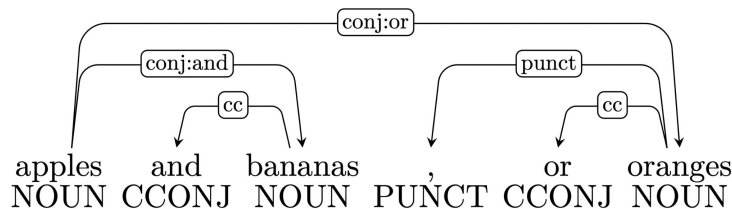


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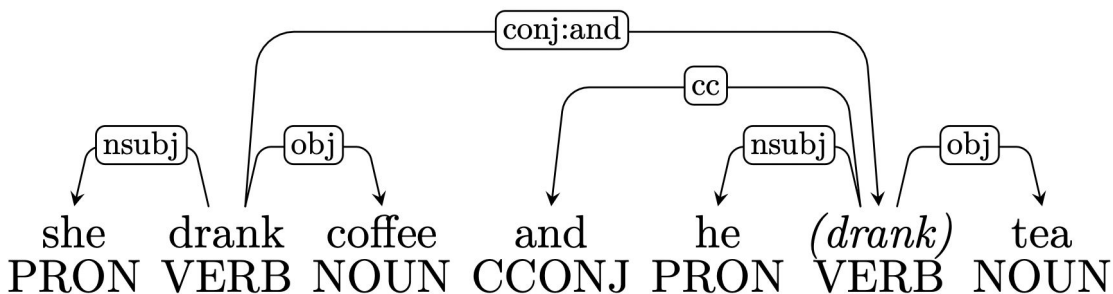


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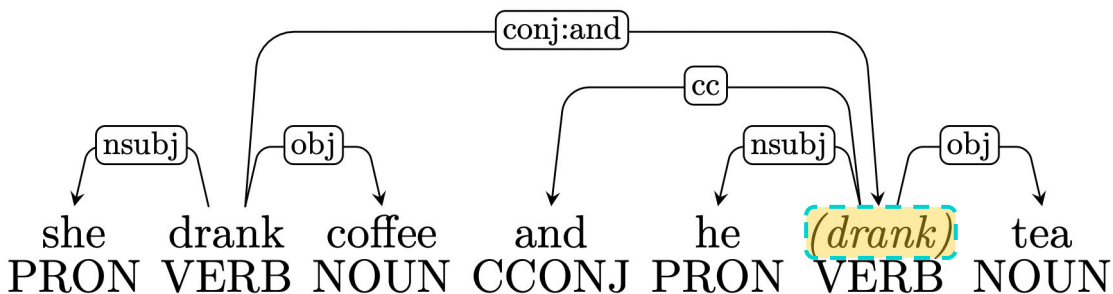
Improved Ellipsis Annotation:



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Improved Ellipsis Annotation:



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Analysis

Employ **two representations of words**
for semantic analysis

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Employ **two representations of words** for semantic analysis

- **WordNet** links words through conceptual relations

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Employ **two representations of words** for semantic analysis

- **WordNet** links words through conceptual relations
- **Word embeddings** represent similarity based on the contexts in which two words appear

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WordNet Relations

Relations used in this project:

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WordNet Relations

Relations used in this project:

- Asymmetric relation:
 - **Hypernymy/hyponymy**

vehicle + car
move + run

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WordNet Relations

Relations used in this project:

- Symmetric relations:
 - **Synonymy**
 - Antonymy
 - Co-hyponymy

car + automobile
friend + ally

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WordNet Relations

Relations used in this project:

- Symmetric relations:
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small + large
wet + dry

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WordNet Relations

Relations used in this project:

- Symmetric relations:
 - Synonymy *red + blue*
 - Antonymy *apple + banana*
 - **Co-hyponymy**

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Word Embeddings

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Word Embeddings

- Google's **Word2Vec** embeddings
- Use **cosine similarity** to test relationship between conjuncts

Part 4: Results and Discussion

- Overview
- WordNet
Analysis
- Word
Embedding
Analysis

Part 4: Results and Discussion

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General Results:

Part 4: Results and Discussion

→ Overview

- WordNet Analysis
- Word Embedding Analysis

General Results:

- 6,892 coordinations in **total**

Part 4: Results and Discussion

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- 6,892 coordinations in **total**
- 6,641 (96.4%) coordinations in **WordNet**

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General Results:

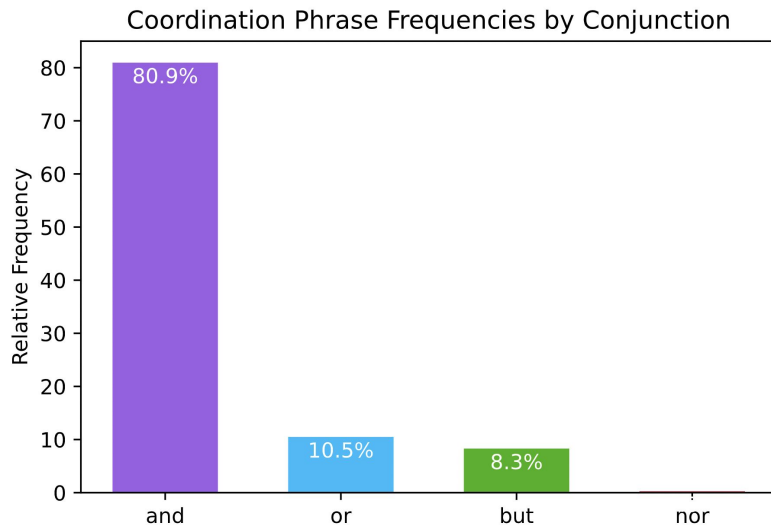
- 6,892 coordinations in **total**
- 6,641 (96.4%) coordinations in **WordNet**
- 27 coordinations (0.4%) include one **elided conjunct**

Part 4: Results and Discussion

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General Results:



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WordNet Analysis by Conjunction:

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WordNet Analysis by Conjunction:

- Examine association between WordNet relations and types of coordinating conjunctions

Part 4: Results and Discussion

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WordNet Analysis by Conjunction:

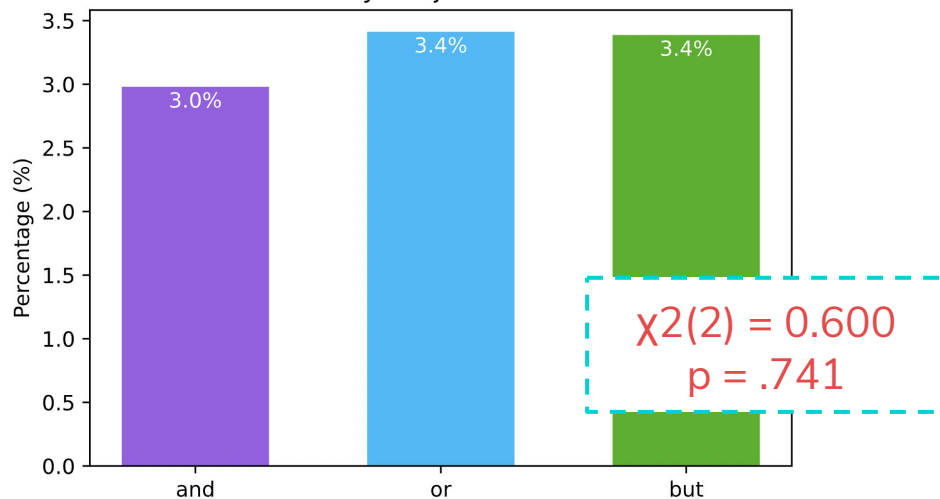
- Examine association between **WordNet relations** and types of **coordinating conjunctions**
- Applies to synonymy, antonymy, and co-hyponymy

Part 4: Results and Discussion

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Synonymy by Conjunction

Percentages of Coordinations with Synonymy
by Conjunction

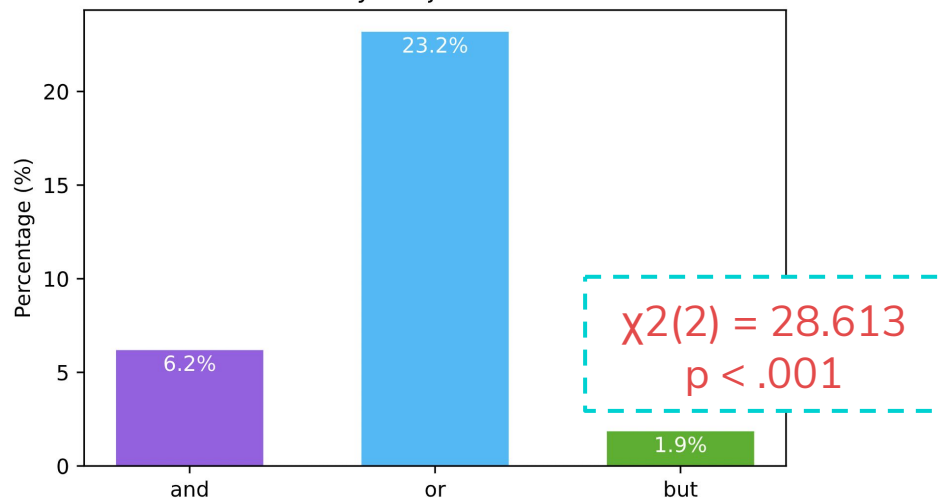


Part 4: Results and Discussion

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Antonymy by Conjunction

Percentages of Coordinations with Antonymy by Conjunction

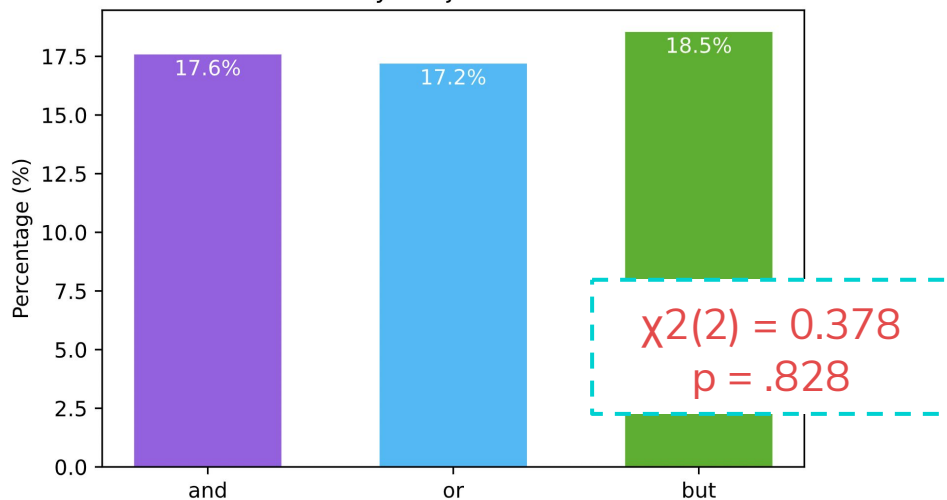


Part 4: Results and Discussion

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Co-hyponymy by Conjunction

Percentages of Coordinations with Co-hyponymy by Conjunction



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This **WordNet Analysis by Conjunction** generally shows:

Part 4: Results and Discussion

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This **WordNet Analysis by Conjunction** generally shows:

- *or*-coordinations have a distinctive use for contrast

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This **WordNet Analysis by Conjunction** generally shows:

- *or*-coordinations have a distinctive use for contrast
- *and* is more general-purpose in use

Part 4: Results and Discussion

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This **WordNet Analysis by Conjunction** generally shows:

- *or*-coordinations have a distinctive use for contrast
- *and* is more general-purpose in use
- *but*-coordinations do not correlate with relations

Part 4: Results and Discussion

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WordNet Analysis by Category:

Part 4: Results and Discussion

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WordNet Analysis by Category:

- Examine association between WordNet relations and the categories of conjuncts

Part 4: Results and Discussion

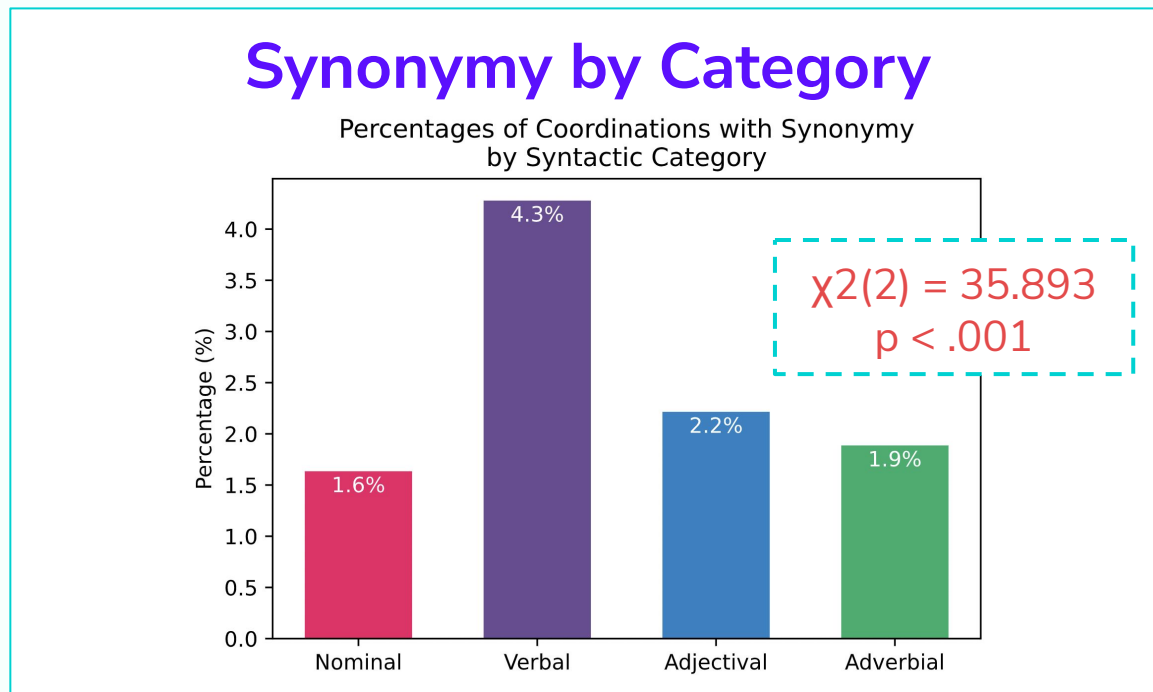
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WordNet Analysis by Category:

- Examine association between WordNet relations and the categories of conjuncts
- Applies to synonymy, antonymy, and co-hyponymy

Part 4: Results and Discussion

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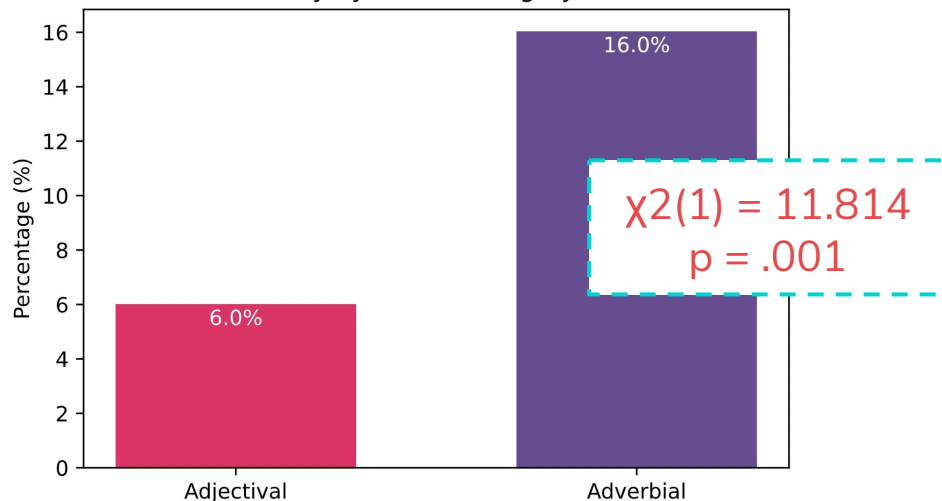


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Antonymy by Category

Percentages of Coordinations with Antonymy
by Syntactic Category

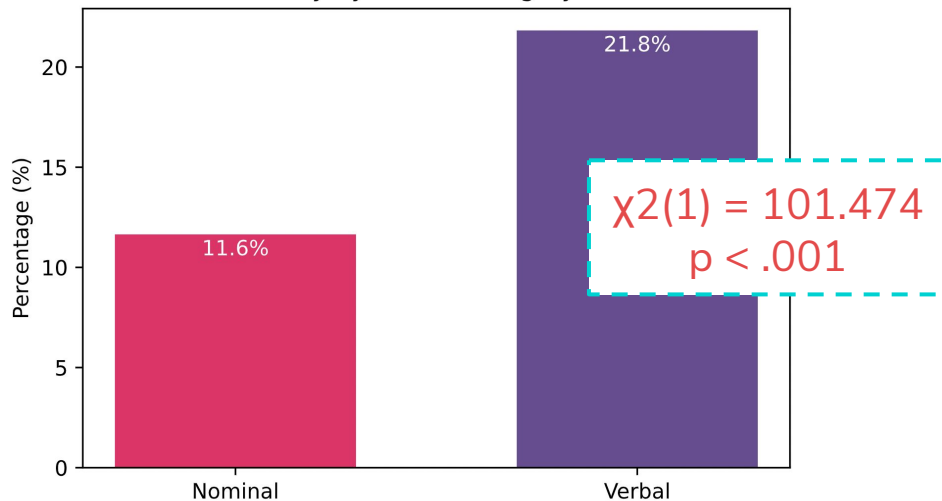


Part 4: Results and Discussion

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Co-hyponymy by Category

Percentages of Coordinations with Co-hyponymy
by Syntactic Category



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This **WordNet Analysis by Category** generally shows:

- Verbal categories correlate with synonymy and co-hyponymy

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This **WordNet Analysis by Category** generally shows:

- Verbal categories correlate with synonymy and co-hyponymy

Trust me, and most especially, [v trust] and [v believe] yourself.

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This **WordNet Analysis by Category** generally shows:

- Antonymous adverbs stem from common phrases

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This **WordNet Analysis by Category** generally shows:

- Antonymous adverbs stem from common phrases
 - “Up and down,” “here or there,” “more or less”

Part 4: Results and Discussion

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Hypernymy

Part 4: Results and Discussion

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Hypernymy

- We consider the **two possible directions** of the relation

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Hypernymy

- We consider the **two possible directions** of the relation
- The **second conjunct** is more often a hypernym of the first conjunct

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Hypernymy

- We consider the **two possible directions** of the relation
- The **second conjunct** is more often a hypernym of the first conjunct
- No significant difference, $\chi^2(1) = 2.045, p = 0.153$

Part 4: Results and Discussion

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Hypernymy

- Hypothesized that hypernymy would apply in certain contexts
I bought [strawberries] and [other fruit].
- Asymmetry is not prominent

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Word Embedding Analysis:

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Word Embedding Analysis:

- Compare the effect of **conjunctions** and **categories** on the **cosine similarity** of conjuncts

Part 4: Results and Discussion

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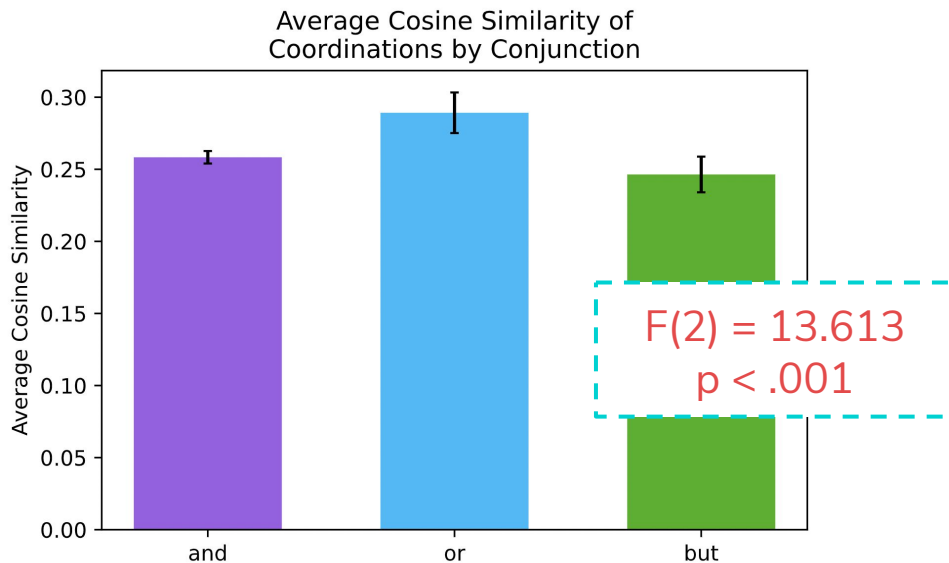
Word Embedding Analysis:

- Compare the effect of **conjunctions** and **categories** on the **cosine similarity** of conjuncts
- Include post-hoc tests

Part 4: Results and Discussion

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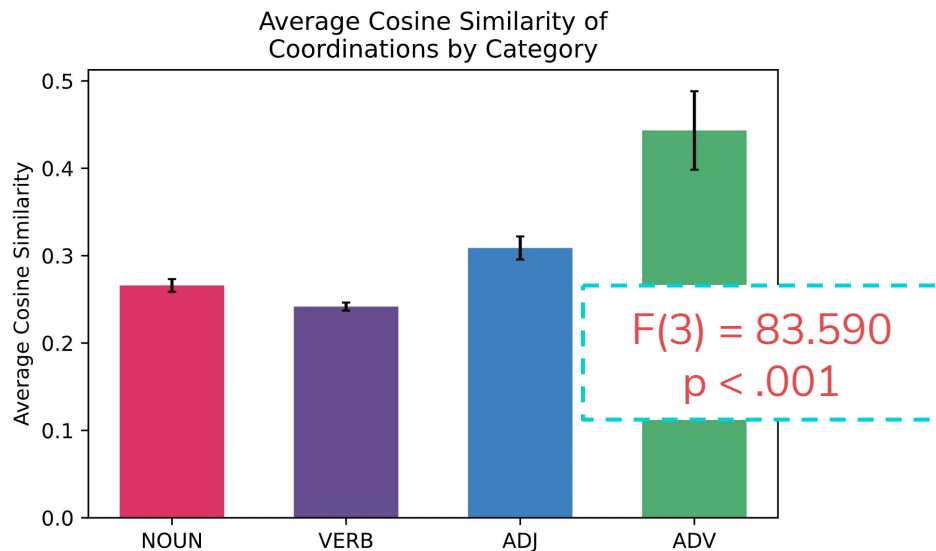
Cosine Similarity by Conjunction



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Cosine Similarity by Category



Part 5: Conclusion

- We use a **computational corpus analysis** to understand semantic patterns in the use of two-termed coordination
- Differences in coordination semantics depend on the **conjunction** and **categories** of the conjuncts
- This work is a step toward a fuller understanding of speakers' **real-world usage** of coordination phrases to better inform linguistic theories

Thank you!

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