# On Commonsense Domains within the Winograd Schema Challenge

#### Aneta Koleva

International Center for Computational Logic Technische Universität Dresden Germany

- Winograd Schema Challenge
- Previous Approaches
- Knowledge Types Identification and Reasoning
- Categorization of Winograd Schemas
- Conclusion

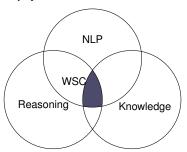


#### **Motivation**

- ▶ Winograd Schema Challenge (Levesque et. al, 2012)
  - S: The trophy does not fit into the brown suitcase because it is too [small/large].
  - Q: What is too [small/large]?
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  - Sentence containing two nouns, one ambiguous pronoun and a special word
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- Characteristics:
  - Easy to answer for an adult English speaker
  - Always contains special word
  - Google proof

#### Competition

- Competition in 2016 at IJCAI-16
  - ▶ Two time-constraint rounds 210 min. each
    - ▶ Pronoun Disambiguation Problems (PDPs) 60
    - Parts of Winograd Schemas 150
  - Four competitors
  - Best result: 58% correctly resolved PDPs
  - There was no second round
- Current state-of-the-art (Radford et. al, 2019) achieves 70.7% accuracy on the WSs dataset

### **Previous Approaches**

- Machine learning and deep learning
  - Supervised ranking SVM
  - Supervised classification Task
  - Knowledge enhanced embeddings
  - Google's language models
  - Open Al language model
- Knowledge-based
  - Knowledge graphs with Relevance Theory
  - ▶ Semantic parsing and knowledge hunting
  - > Parsing query results and assigning scores
  - Knowledge types identification

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- Language models assign probabilities to both sentences
- Evaluation and results
  - PDPs 70% accuracy
  - ▶ WSC 63.7% accuracy

# Knowledge Types Identification and Reasoning (Sharma and Baral, 2018)

- ▶ Identified 12 knowledge types which cover the entire WSC dataset
- Developed a logical reasoning algorithm
- ► Evaluated on 100 problems from WSC and achieved 100% accuracy

<sup>1</sup>kparser.org

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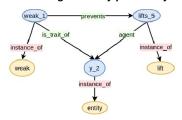
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- Solver
  - 1. Semantic graph<sup>1</sup> of the input sentence and question
  - 2. Semantic graph representation of background knowledge
  - 3. Graph merging
  - 4. Project question graph on the merged graph
  - 5. Answer the node from the merged graph which is from the same domain as the unknown node from the question graph



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► Representation of the knowledge "weak y prevents y lifts"





### Methodology

### **Categorization of Winograd Schemas**

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- ▶ Idea
  - Analyze the input Winograd Schema and identify the domain
  - 2. Search for knowledge specific to this domain
  - 3. Apply reasoning procedure

### **Identified Categories**

Category	Example	
Physical	S: John couldn't see the stage with Billy in front of him because he is so [short/tall].	
	Q: Who is so [short/tall]?	
2. Emotions	S: Frank felt [vindicated/crushed] when his longtime rival Bill	
	revealed that he was the winner of the competition.	
	Q: Who was the winner of the competition?	
3. Interactions	S: Joan made sure to thank Susan for all the help she had [given/received].	
	Q: Who had [given/received] help?	
4. Comparison	S: Joe's uncle can still beat him at tennis, even though he is 30 years [older/younger].	
	Q: Who is [older/younger]?	
5. Causal	S: Pete envies Martin [because/although] he is very successful.	
	Q: Who is very successful?	
Multiple knowledge	S: Sam and Amy are passionately in love, but Amy's parents are unhappy about it,	
	because they are [snobs/fifteen].	
	Q: Who are [snobs/fifteen]?	

### **Annotation of Winograd Schemas**

- Strong agreement between the annotators Cohen's kappa score 0.66
- Annotation Results

Category	Annotator 1	Annotator 2
Physical	36	39
Emotions	7	9
Interactions	44	24
Comparison	19	26
Causal	16	18
Multiple knowledge	28	34

### **Graph Representation for Physical Category**

- 1. The man couldn't lift his son because he was so weak.
- 2. The trophy doesn't fit into the brown suitcase because it's too small.

### Reasoning

- Knowledge required for both examples is about physical features
- Similar reasoning rules for categorizing the traits
  - 1. weak(X) :- lift(X,Y), not lift(modifier, could).
  - 2. small(Y) :- fit(X,Y), not fit(modifier, could).

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- Reasoning Algorithm
- Change of background knowledge
  - has\_k(weak,prevents,lift).

#### **Contributions**

- Overview of different approaches towards WSC
- None achieves close to 90% accuracy
- We analyzed the entire WSC corpus and identified 6 categories
- We identified a mistake in the Reasoning Algorithm and proposed a correction

#### **Future Work**

- ▶ Better Reasoning Algorithm
- ► Knowledge Graphs (RDF) representation