

# Abstract Meaning Representation of Constructions: The More We Include, the Better the Representation

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LREC

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Where does meaning come from?

Individual words compose meaning

Lexical Predicate

She moved the foam off her cappuccino NP. Agent NP. Theme PP. Path

 Flexible templates (compatible with certain words) can also carry meaning

Construction;

Caused-Motion She moved the foam off her cappuccino

NP. Agent Verb NP. Theme PP. Path



Where does meaning come from? Why does this matter?

### **NLP Impact:**

- What do we store in a computational lexicon?
- Semantic Role Labeling / Syntactic Parsing: What do we assume are predicates and arguments of those predicates?



What do we store in a computational lexicon? What do I consider predicates and their args?

Individual words

Lexical Predicate

She moved the foam off her cappuccino NP. Agent NP. Theme PP. Path

Constructions (pairing of form + meaning)

Construction: She mode Caused-Motion NP. Agent V

She moved the foam off her cappuccino

NP. Agent Verb NP. Theme PP. Path



What do we store in a computational lexicon? What do I consider predicates and their args?

Individual words

Lexical Predicate

She moved the foam off her cappuccino NP. Agent NP. Theme PP. Path

Constructions (pairing of form + meaning)

Construction: Caused-Motion

She sneezed the foam off her cappuccino

NP. Agent Verb NP. Theme PP. Path



# Background: Constructions

She sneezed the foam off her cappuccino.

- Sneeze.01 (typically intransitive)
  - Arg0: sneezer
- Caused Motion Construction
  - Mover, moved, path

Argument Structure Constructions: productive patterns, licensing verb and arguments









Argument Structure Constructions: Goldberg, 1995

## Research Problem

How can we extend the Abstract Meaning Representation (AMR) to account for meaning stemming from constructions?



# Background: AMR

- Goals:
  - creating large-scale semantics bank
  - simple structures, like Penn Treebank
- Supporting research in:
  - semantic parsing
  - natural language generation
  - machine translation
  - 70 plus research papers use AMR!



# Background: AMR

AMR assigns semantic roles of individual lexical predicates.

- Assign.01 from PropBank "Rolesets"
  - ARG0 (assigner): AMR
  - ARG1 (assigned): semantic roles
  - ARG2 (assigned-to): individual lexical predicates



PropBank: Palmer et al., 2005; http://propbank.github.io

# Background: AMR

AMR assigns semantic roles...
AMR assignment of semantic roles of individual lexical predicates...

should represent concepts and relations consistently, despite syntactic differences.

- Assignment → Assign.01
  - ARG0 (assigner): AMR
  - ARG1 (assigned): semantic roles
  - ARG2 (assigned-to): individual lexical predicates



## AMR Approach to Constructions

The more we include, the better the representation.

- Include.01, representation → represent.01,
   better → good.02
- Gap in representation: Correlation

Annotating constructions required a novel approach...



## AMR Approach to Constructions

- 1. Exploiting lexical predicate rolesets in combination with modifier roles (e.g., Source, Destination), addition of implicit predicates (e.g., Cause-01, Move-01)
  - Where existing AMR machinery provides adequate coverage of constructional meaning
- 2. Adding constructional rolesets
  - Where existing AMR machinery does not adequately capture semantics, and/or
  - We can add a single construction roleset in lieu of many individual lexical rolesets



# Exploiting Lexical Rolesets

• Intransitive Motion Construction:

```
Rumble-01
Arg0: entity rumbling
Arg1: sound/utterance
```

2. The troops rumbled along the main road.

Caused-Motion Construction:

```
Blink-01
Arg0: blinker
Arg1: eyes (usually unstated)

3. He blinked the snow off his eyelashes.
(b / blink-01
:ARG0 (h/ he))
:ARG0-of (c5 / cause-01)
:ARG1 (m2 move-01)
:ARG1 (s / snow)
:source (e / eyelash
:part-of h))))
i.e. He blinked, the blinking caused the snow to move from
```



Arg2: hearer

his eyelashes.

## Adding Constructional Rolesets

- Degree-Related Constructions Have-Degree-91:
  - Comparison
  - Superlative
  - Degree-consequence
- Quantity-Related Constructions Have-Quant-91:
  - Comparison
  - Superlative
  - Quantity-consequence
- The X-er, The Y-er Correlate-91
- Comparing Resemblance Have-Degree-of-Resemblance-91



Construction lexicon: FrameNet Constructicon, Fillmore et al. 2012

## Degree-Related Constructions

#### Have-Degree-91

```
Arg1: domain, entity characterized by attribute
```

Arg2: attribute (e.g. tall)

Arg3: degree itself (e.g. more/most, less/least, equal)

Arg4: compared-to

Arg5: superlative: reference to superset

Arg6: consequence, result of degree

#### Comparative:

```
4. The girl is taller than the boy.
```

```
(h / have-degree-91
```

:ARG1 (g / girl)

:ARG2 (t / tall)

:ARG3 (m / more)

:ARG4 (b / boy))

i.e. *The girl is more tall compared to the boy.* 

#### Superlative:

```
5. She is the tallest girl on the team.
```

```
(h / have-degree-91
     :ARG1 (s / she)
     :ARG2 (t / tall)
```

:ARG3 (m / most)

:ARG5 /g / girl

:ARG0-of (h2 / have-org-role-91

·ARG1 (t2 / team)))

i.e. She is the most tall of the girls on the team.



## Degree-Related Constructions

#### **Have-Degree-91**

Arg1: domain, entity characterized by attribute

Arg2: attribute (e.g. tall)

Arg3: degree itself (e.g. more/most, less/least, equal)

Arg4: compared-to

Arg5: superlative: reference to superset

Arg6: consequence, result of degree

Degree-Consequence: The watch is too wide; therefore, it does not fit my wrist.

I was too tired to drive.



## The X-er, The Y-er

#### Correlate-91

Arg1: X, degree/quant word modifying first item changing in relation to Arg2

Arg2: Y, degree/quant word modifying second item changing

in relation to Arg1

10. The longer he is around, the more miserable I will be.

```
(c / correlate-<u>91</u>
        :ARG1 (m2 / more
              :ARG3-of (h2 / have-degree-91
                     :ARG1 (b / be-located-at-91
                           :ARG1 (h / he)
                           :ARG2 (a / around))
                     :ARG2 (l2 / long-03
                           :ARG1 b)))
        :ARG2 (m3 / more
              :ARG3-of (h3 / have-degree-91
                    :ARG1 (i / i)
                     :ARG2 (m / miserable))))
```

i.e. An increase in how long he is around correlates with an increase in how miserable I am.



## Evaluation, Implementation

- New guidelines, rolesets piloted on 'Challenge Set'
  - 50 sentences from AMR 2.0
  - Selected using keyword searches, manual analysis
  - Represents variety of degree/quantity related constructions
  - Includes tricky cases with clear inconsistencies in past annotation
- Double annotated: 1 CU annotator, 1 SDL annotator
- Agreement: 88.6% ('smatch' score (Cai and Knight, 2013))
- Manual retrofitting of approximately 4700 annotations



## Conclusions, Future Work

- AMR 3.0 release 2018
  - 59783 total AMRs
  - 6112 instances of degree/quantity-based constructions
- Coverage of constructional semantics: a layer of meaning critical for translation, natural language understanding
  - 4 construction entries added to the AMR lexicon
  - 5 distinct constructions
- Deepening AMR...
  - More constructions?
  - Aspect, Modality

N / [ _ ]	ti-sentence
- IVI I I I	TI-CENTENCE
	d schiche

Use Case	Roleset/Relation	Count
Downtoners, in-	Degree	4547
tensifiers		
Comparison, su-	Have-Degree-	4943
perlative, degree-	91	
consequence		
Comparison,	Have-Quant-91	1122
superlative,		
quantity-		
consequence,		
quantity reifica-		
tion		
Comparing	Have-	9
resemblances	Degree-of-	
	Resemblance-	
	91	
The X-er, The Y-	Correlate-91	38
er		

# thank you



## Collaborators





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Tim O'Gorman





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# Background: Constructions

Alternative: Additional senses of lexical predicates (e.g., caused-motion sense of sneeze)

The child ???? her foot out of the boot.

Gary talked me into a corner.

They booed the clown off the stage.

She blinked the snow off her eyelashes.

Caused Motion: She sneezed the foam off her cappuccino

Syntax: NP V NP PP

Semantics: Agent V Theme Initial Location



## Research Problem

Where does meaning come from?

tall, modifier *She is as tall as her brother.* 

**Lexical Semantics**pull They pulled the clown off the stage.

motion boo They booed the clown off the stage.

They booed the clown off the stage.

Caused-Motion

They booed the snow off his eyelashes.

Caused-Motion

adverbial, sell *The lower the price, the more you'll sell*. Correlation

Comparison

To be comprehensive, Abstract Meaning Representation must include both lexical, constructional semantics



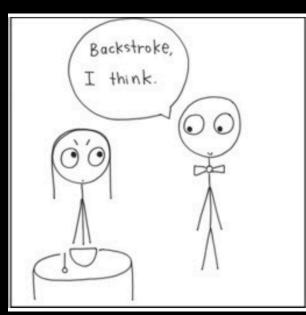
## Background: Constructions

Constructions: prefabricated parts, templates; pairing of form and meaning arising out of individual discourse experience.



Compositional: WH-Question

Constructional: Surprise, Disapproval



Construction Grammar: Hopper, 1998; MacWhinney, 2001; Bybee and McClelland, 2005; Fillmore et al., 1988; Kay and Fillmore, 1999; Michaelis and Lambrecht, 1996.

