The BECauSE Corpus 2.0: Annotating Causality and Overlapping Relations

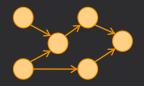
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Recognizing causal assertions is critical to language understanding.

Ubiquitous in our mental models



Ubiquitous in language

12% of explicit discourse connectives in Penn Discourse Treebank (Prasad et al., 2008)

Useful for downstream applications (e.g., information extraction)

The prevention of FOXP3 expression was not caused by interferences.

BECauSE draws on ideas from Construction Grammar (CxG) to annotate a wide variety of causal language.

Such swelling can impede breathing.

(Verbal)

They moved **because of** the schools.

(Prepositional)

Our success is **contingent on** your support.

(Adjectival)

We're running late, so let's move quickly.

(Conjunctive)

This opens the way for broader regulation.

(Multi-word expr.)

For markets to work, banks can't expect bailouts.

(Complex)

Causal language is difficult to disentangle from overlapping semantic domains.

After a drink, she felt much better.

(Temporal)

They're too big to fail.

(Extremity)

The more I read his work, the less I like it.

(Correlation)

The police **let** his sister visit him briefly.

(Permission)

As voters get to know Mr. Romney, his poll numbers will rise.

(Temporal + Correlation)

Main contributions of this paper:

- The BECauSE 2.0 annotation scheme including 7 overlapping relation types
- 2. The updated & expanded BECauSE 2.0 corpus
- Evidence about how meanings compete for linguistic machinery

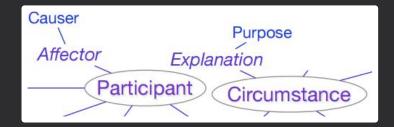
Several general-purpose schemes include some elements of causal language.

PropBank, VerbNet

(Palmer et al., 2005; Schuler, 2005) Roleset id: prevent.01, stop, prevent, stopping in advance

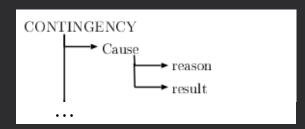
Prepositions

(Schneider et al., 2015, 2016)



Penn Discourse Treebank

(Prasad et al., 2008)



FrameNet |

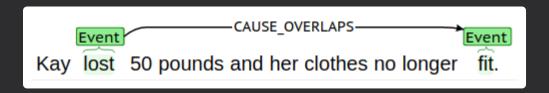
(Ruppenhofer et al., 2016)



Others have focused specifically on causality.

CaTeRS

(Mostafazadeh et al., 2016)



Richer Event Description

(O'Gorman et al., 2016)

BEFORE-PRECONDITIONS

We've allocated a budget to equip the barrier with electronic detention equipment.

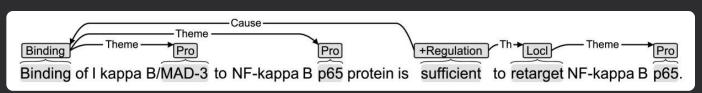
Causality in TempEval-3

(Mirza et al., 2014)



BioCause

(Mihaila et al., 2013)



BECauSE 1.0 annotates causal language, expressed using arbitrary constructions.

Bank of Effects and Causes Stated Explicitly

Such swelling can **impede** breathing.

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 - i. Practices retained from BECauSE 1.0
 - ii. Improvements and extensions in BECauSE 2.0

Causal language:
a clause or phrase in which
one event, state, action, or entity
is explicitly presented
as promoting or hindering
another

Connective: fixed lexical cue indicating a causal construction

John killed the dog because it was threatening his chickens.

John prevented the dog from eating his chickens.

Ice cream consumption causes drowning.

She must have met him before, because she recognized him yesterday.

Not "truly" causal

Effect: presented as outcome Cause: presented as producing effect

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Annotators were guided by a "constructicon."

| Connective pattern | <cause> prevents <effect> from <effect></effect></effect></cause> | <enough cause=""> for <effect></effect></enough> |
|------------------------|---|--|
| Annotatable words | prevent, from | enough, for, to |
| WordNet verb senses | prevent.verb.01 prevent.verb.02 | |
| Туре | Verbal | Complex |
| Degree | Inhibit | FACILITATE |
| Type restrictions | Not Purpose | |
| Example | His actions prevented disaster. | There's enough time for you to find a restroom. |

Causation can be positive or negative.

This has often **caused** problems elsewhere.



He **kept** the dog **from** leaping at her.



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Update I:Three types of causation

The system failed because of a loose screw.



Mary left because John was coming.



Mary left in order to avoid John.



The engine is still warm, so it must have been driven recently.



Update 2: Means arguments for cases with an agent and an action

My dad caused a commotion by shattering a glass.

Cause Effect Means

By altering immune responses, inflammation can trigger depression.

Update 3: Overlapping semantic relations are annotated when they can be coerced to causal interpretations.

```
After last year's fiasco, everyone is being cautious.

MOTIVATION ARGC ARGE

+ TEMPORAL
```

```
After last year's fiasco, they've rebounded this year.

TEMPORAL ARGC ARGE
```

He won't be back until after Thanksgiving.

We annotate 7 different types of overlapping relations.

TEMPORAL After; once; during

CORRELATION As; the more...the more...

HYPOTHETICAL If...then...

Obligation/Permission Require; permit

CREATION/TERMINATION Generate; eliminate

EXTREMITY/SUFFICIENCY So...that...; sufficient...to...

CONTEXT Without; when (non-temporal)

Annotators applied several tests to determine when an overlapping relation was also causal.

- Can the reader answer a "why" question?
- Does the cause precede the effect?
- Counterfactuality: would the effect have been just as probable without the cause?
- Ontological asymmetry: could the cause and effect be reversed?
- Can it be rephrased as "because?"

(see Grivaz, 2010)

Inter-annotator agreement remains high.

| | Causal | Overlapping | |
|-----------------------------|--------|-------------|--|
| Connective spans (F_1) | 0.77 | 0.89 | |
| Relation types (κ) | 0.70 | 0.91 | |
| Degrees (κ) | 0.92 | (n/a) | |
| CAUSE/ARGC spans (%) | 0.89 | 0.96 | |
| CAUSE/ARGC spans (Jaccard) | 0.92 | 0.97 | |
| CAUSE/ARGC heads (%) | 0.92 | 0.96 | |
| EFFECT/ARGE spans (%) | 0.86 | 0.84 | |
| EFFECT/ARGE spans (Jaccard) | 0.93 | 0.92 | |
| EFFECT/ARGE heads (%) | 0.95 | 0.89 | |

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- 2. The updated & expanded BECauSE 2.0 corpus
- 3. Evidence about how meanings compete for linguistic machinery

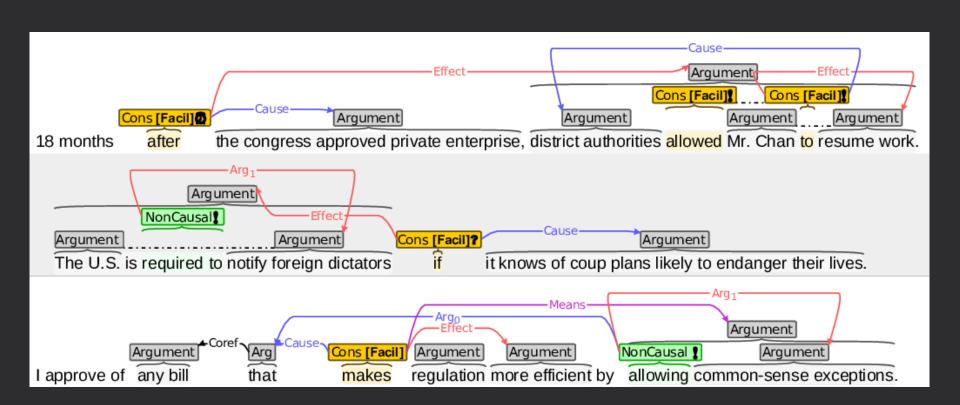
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We have annotated an augmented corpus with this scheme.

| | Documents | Sentences | Causal | Overlapping |
|--|-----------|-----------|--------|-------------|
| New York Times Washington section (Sandhaus, 2014) | 59 | 1924 | 717 | 519 |
| Penn TreeBank WSJ | 47 | 1542 | 534 | 340 |
| 2014 NLP Unshared Task in PoliInformatics (Smith et al., 2014) | 3 | 695 | 326 | 149 |
| Manually Annotated Sub-Corpus (Ide et al., 2010) | 12 | 629 | 228 | 166 |
| Total | 121 | 4790 | 1805 | 1174 |

bit.ly/BECauSE

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Causality has thoroughly seeped into the temporal and hypothetical domains.

Of the causal expressions in the corpus:

- > 14% are piggybacked on temporal relations
 - ~7% are expressed as hypotheticals

Conditional hypotheticals don't have to be causal, but most are.

Non-causal: If he comes, he'll bring his wife.

Causal: If I told you, I'd have to kill you.

84% carry causal meaning

We seem to prefer describing causation in terms of agents' motivations.

~45% of causal instances are MOTIVATION or PURPOSE

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Lingering difficulties include other overlapping relations and bidirectional relationships.

Origin/destination toward that goal

Topic fuming over recent media reports

Component as part of the liquidation

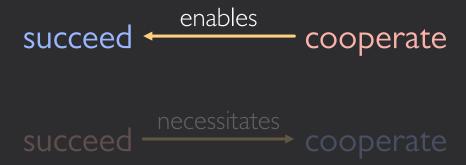
Evidentiary basis went to war **on** bad intelligence

Having a role as an American citizen

Placing in a puts us at risk position

Lingering difficulties include other overlapping relations and bidirectional relationships.

For us to succeed, we all have to cooperate.



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