

Modeling Linguistic Structure (And What We Can Do With One)

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School of Computing



Joint work with many people at Universities of Illinois, Colorado, Georgetown University,
Stanford University and IHMC Florida

This talk

- The Ins and Outs of Preposition Semantics
- Predicting Preposition Relations
- Learning to Read

What are structures and why are they useful?

- For the purpose of this talk
 - A labeled graph
- Computer science knows how to deal with discrete objects
 - Databases, queries, graph algorithms, etc
- But, natural language is *unstructured*

The agenda: Convert unstructured text into linguistically motivated structured representations

Discrete Linguistic Structures

- Frame Semantics, case grammars
 - Rich history in linguistics starting with Fillmore (1966)
- Large datasets to train classifiers
 - PropBank [Palmer, et al 2005-]
 - FrameNet [Fillmore, Ruppenhofer,...]
- The general premise: Words evoke frames
 - Which words? Most commonly, verbs and eventive nouns

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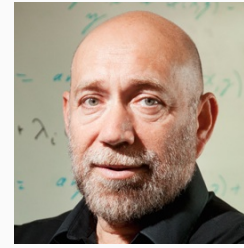
The Ins and Outs of Preposition Semantics



Nathan Schneider



Jena Hwang



Dan Roth



Meredith Green



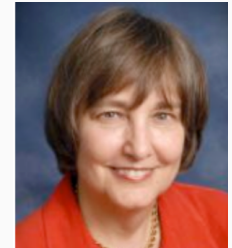
Abhijit Suresh



Tim O’Gorman

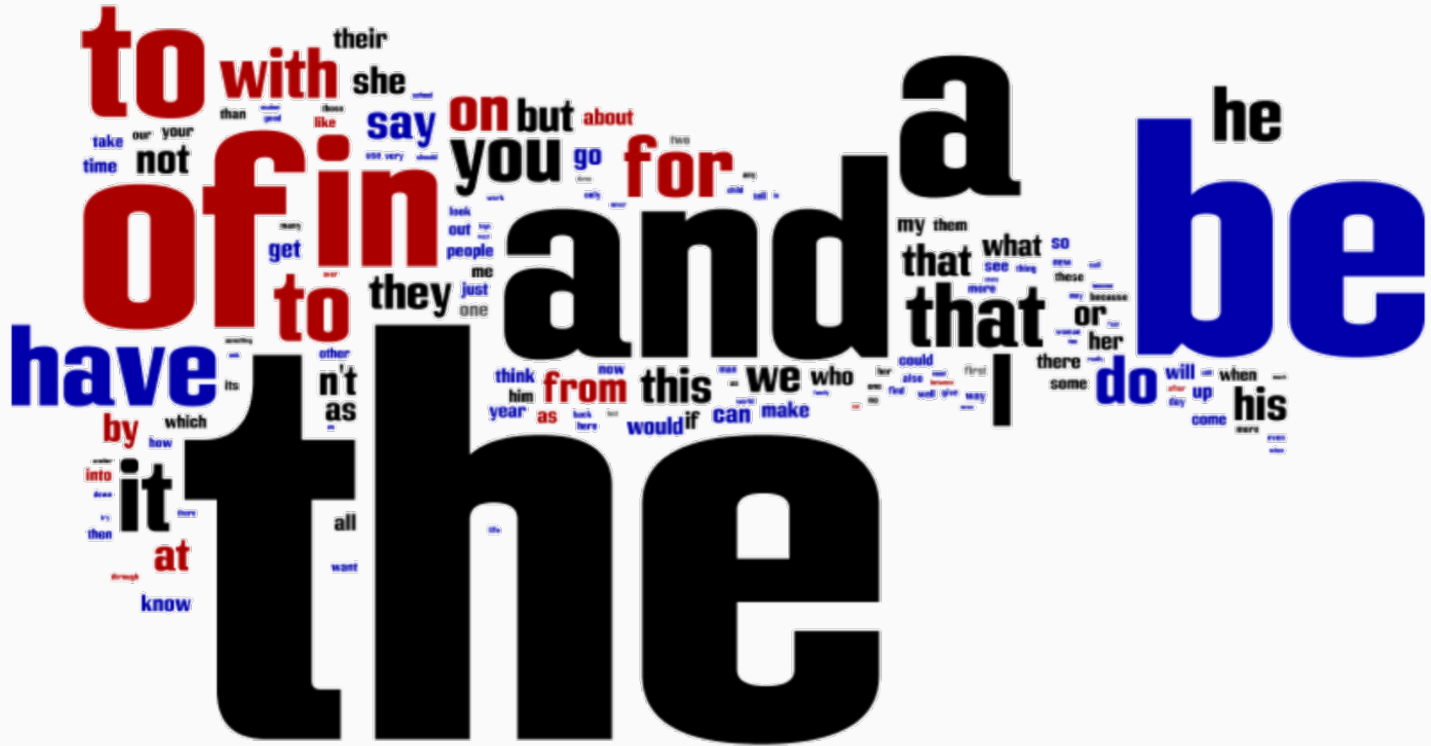


Katie Conger



Martha Palmer

Prepositions abound in English



based on COCA list of 5000 most frequent English words

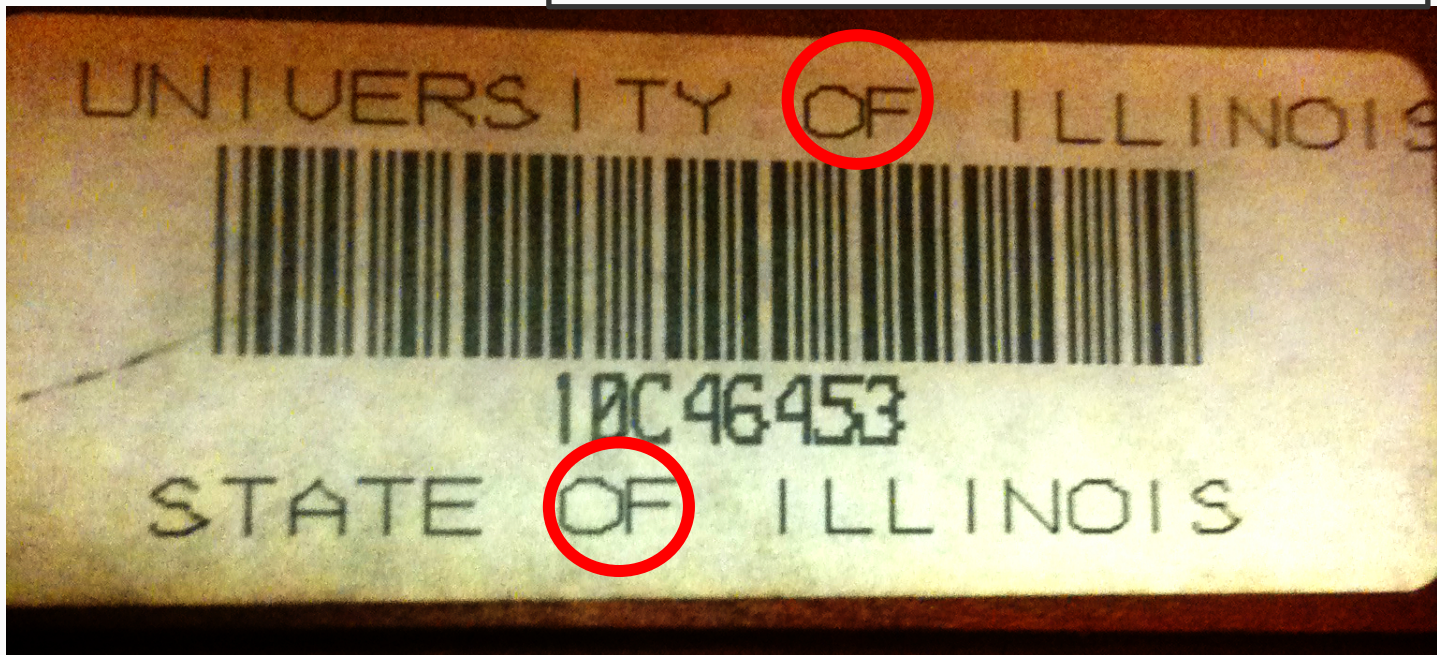
Prepositions trigger semantic relations



Due to his grammar mistake, Wilbur found a position. It just wasn't the one he wanted.

Prepositions are ambiguous

The university belongs to Illinois



Illinois is a state

Preposition Sense Disambiguation

Eg. *State **of** Illinois* vs. *University **of** Illinois*

- The Preposition Project [Litkowski and Hargraves, 2005]
 - Word sense for 34 prepositions
 - Based on preposition definitions in Oxford Dictionary of English

Preposition Super senses

Merriam-Webster

over

Fine grained details

Like Over 15 (6) – 1

Full Definition of OVER

- 1** —used as a function word to indicate motion or situation in a position higher than or above another <towered *over* his mother> <flew *over* the lake> <rode *over* the old Roman road>
- 2** **a** —used as a function word to indicate the possession of authority, power, or jurisdiction in regard to some thing or person <respected those *over* him>
b —used as a function word to indicate superiority, advantage, or preference <a big lead *over* the others>
c —used as a function word to indicate one that is overcome, circumvented, or disregarded <passed *over* the governor's veto>
- 3** **a** : more than <cost *over* \$5>
b : ABOVE 4
- 4** **a** —used as a function word to indicate position upon or movement down upon <laid a blanket *over* the child> <hit

Supersenses = shared functions

They ran **to** the roof **for** a quick escape.



Destination



Purpose



They ran **for** the roof **to** escape the cops.

Adposition Supersenses

- **Early attempt: Preposition relations**
 - Mapping preposition senses to unlexical supersenses [TACL 2013]
- **English preposition supersenses**
 - Annotation scheme & lexicon [LAW 2015]
 - Annotated corpus [LAW 2016]
- **CARMLS**: making these categories multilingual [work in progress!]

English Preposition Supersenses data

Comprehensive annotation: first dataset with **all** prepositions (types+tokens) semantically annotated

- Sentences not hand-selected
- Sentences fully annotated
- Preposition types not constrained by a lexicon (labels generalize)
- All sentences seen by multiple annotators

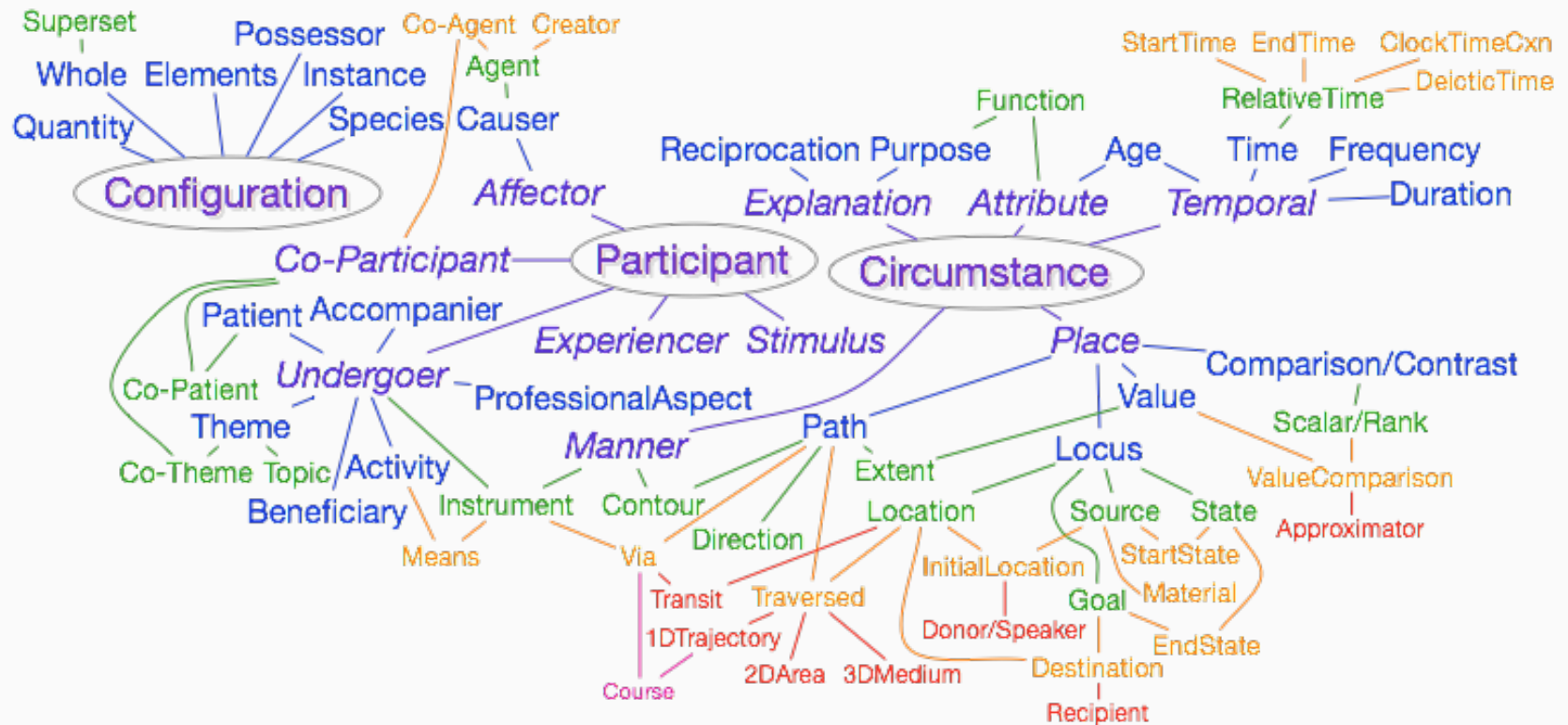
Challenges

1. Scoping phenomena of interest: multiwords, particles, grammaticized uses of prepositions, etc.
2. Deciding categories and their boundaries
3. Thorough documentation with example usages

Defining the supersenses

- Starting point: 32 categories from S & Roth (2013) + VerbNet thematic role hierarchy
- Extensive discussion of examples and pilot annotation to identify categories
 - Multiple inheritance among categories
E.g., **VIA** (*talk over Skype*) = **INSTRUMENT** + **PATH**

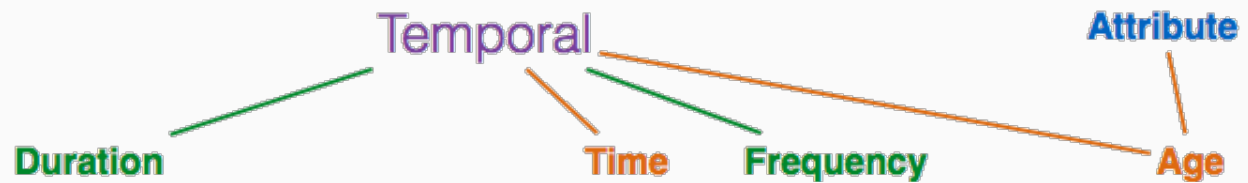
Supersense hierarchy



75 preposition supersense categories

<http://tiny.cc/prepwiki>

Temporal hierarchy



PrepWiki

- Because there are so many supersenses, we showed annotators **example sentences** for each prep/supersense pair
 - So we needed to create a lexical resource documenting possible usages

Category:In

<http://tiny.cc/prepwiki>

Word sense

	Definition	Example	Supersense
In 1(1)	expressing the situation of something that is or appears to be enclosed or surrounded by something else:	I'm living in London; she saw the bus in the rear-view mirror.; examples	SST-Location
In 1(1)-1	used to introduce someone's clothing	Trouserless men looked absurd in socks.	SST
In 10(7a)	indicating the key in which a piece of music is written:	Mozart's Piano Concerto in E flat.;	SST-Attribute
In 11(8)	as an integral part of (an activity):	In announcing the program, Computershare pointed out the environmental benefits of reducing the use of valuable resources such as trees.; He was unsuccessful in running for president	SST-Circumstance
In 12(9)	expressing a value as a proportion of (a whole):	a local income tax running at six pence in the pound.; Perhaps only one in twenty of the city's adult residents had been born there.	SST-Superset, Value
In 2(1a)	expressing motion with the result that something ends up within or surrounded by something else:	don't put coal in the bath; he got in his car and drove off.;	SST-Location
In 3(2)	expressing a period of time during which an event happens or a situation	they met in 1885; at one o'clock in the morning; I hadn't seen him in years.; In my life I have never seen so many cats on a couch; the leaders met for	SST-DeicticTime, Time

supersense(s)
mapped to individual examples
(not shown)

Corpus annotation

- Fully annotated a 55,000 word corpus of online reviews
- Several annotators at CU-Boulder
- All sentences were independently annotated twice, then adjudicated by an expert annotator
 - Original IAA rates varied considerably; mostly 60%–78%
 - IAA between experts: 88%

Preposition types

to
of
in
for
with
on
at

from
about
like
by
after
as
back

before
over
than
since
around
into
out

without
off
until
through
ago
away
within

during
...

out of
at least
due to
because of
at all
less than

as soon as
all over
instead of
other than
next to
nothing but

in front of
more than
more like
along with
just about
thanks to

rather than
as to
such as
as long as
aside from
...

Ongoing work: CARMLS

Case and Adposition
Representation
for Multi-Lingual Semantics

Can we use supersenses for case markers and adpositions in other languages?

Focus languages (besides English)

Hindi

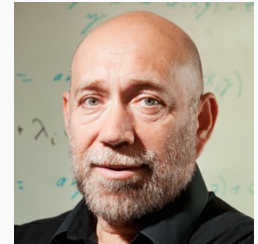


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la-Rae Han

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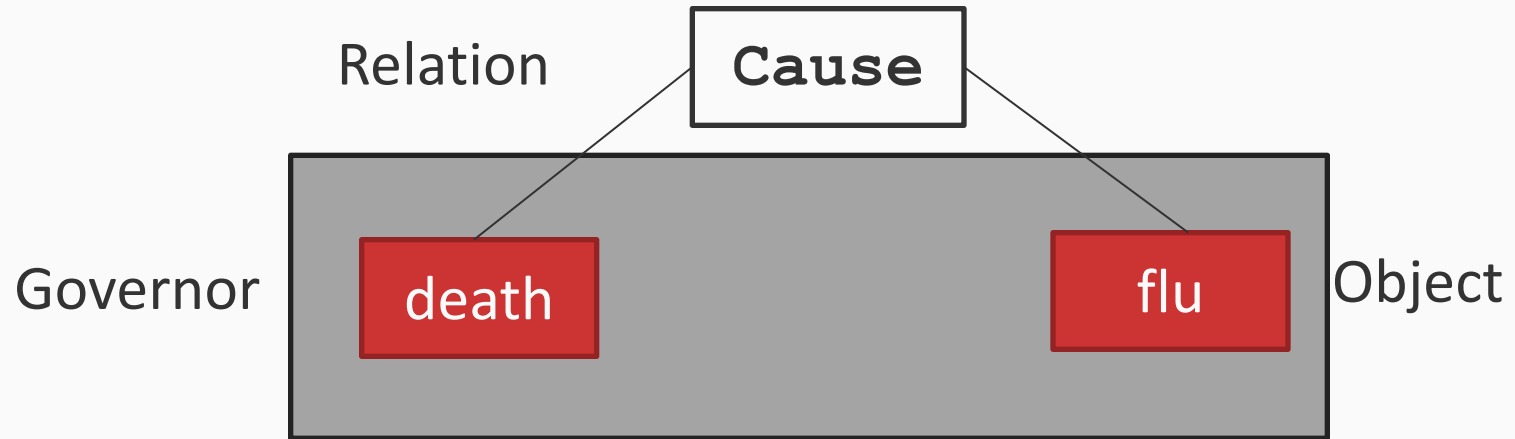
Dan Roth

“zoo in NYC” \longrightarrow **Location**(zoo, NYC)

Two Models for Predicting Preposition Relations [TACL 2013]

Structure of prepositions

Poor care led to her death **from** flu.



Relation depends on argument types

Poor care led to her death **from** flu.

Cause(death, flu)

Poor care led to her death **from** pneumonia.

How do we generalize the classifier to unseen arguments in the same “type”?

Why are types important?

- Goes beyond words
 - Abstract *flu* and *pneumonia* into the same group
- Some semantic relations hold only for certain types of entities
- How do we know the types of words?
A: Hypernymy

WordNet IS-A hierarchy

pneumonia

=> respiratory disease

=> disease

=> illness

=> ill health

=> pathological state

=> physical condition

=> condition

=> state

=> attribute

=> abstraction

=> entity

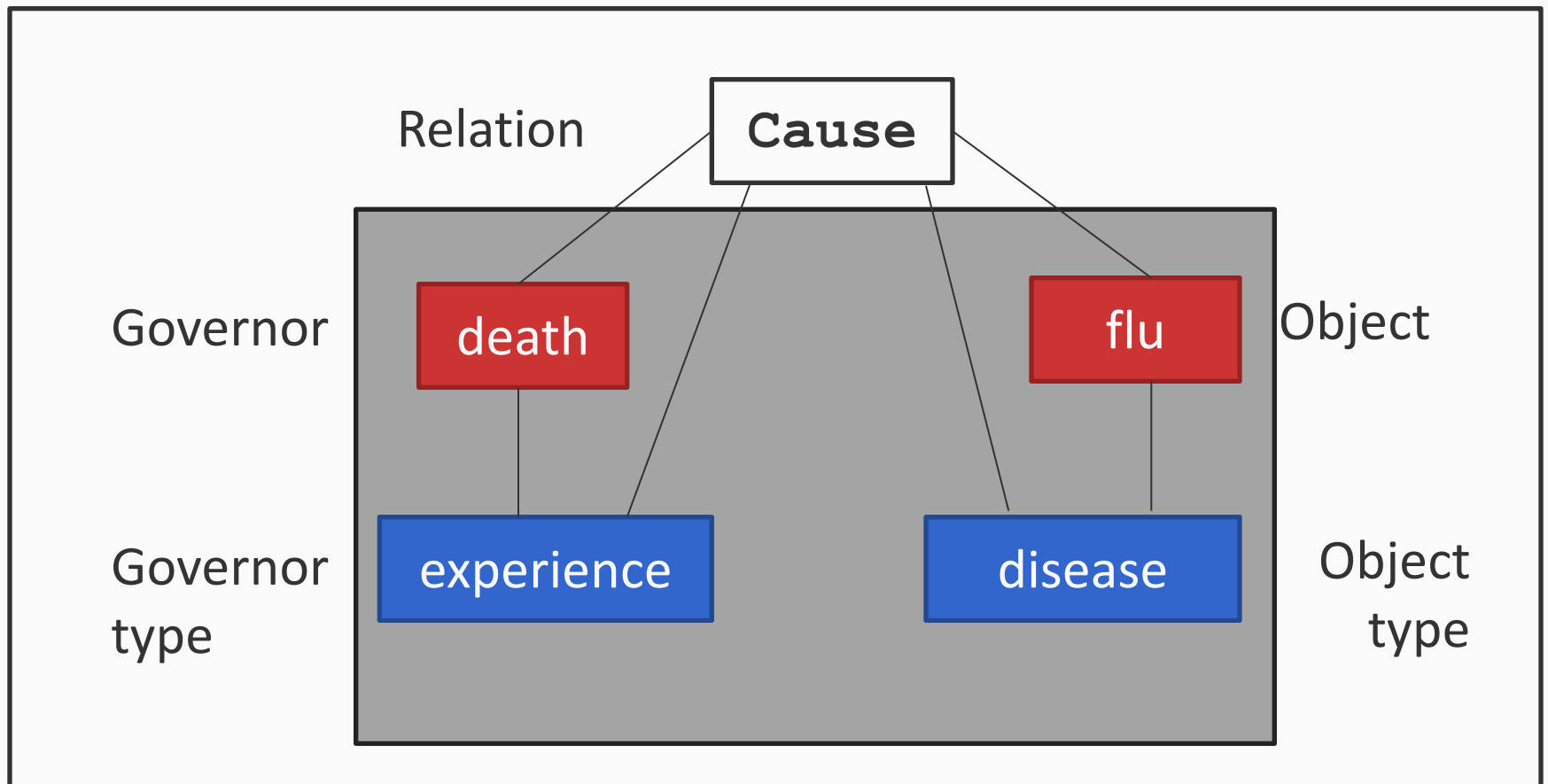
Picking the right level in this hierarchy can generalize pneumonia and flu

More general, but less discriminative

Picking incorrectly will over-generalize

Structure of prepositions

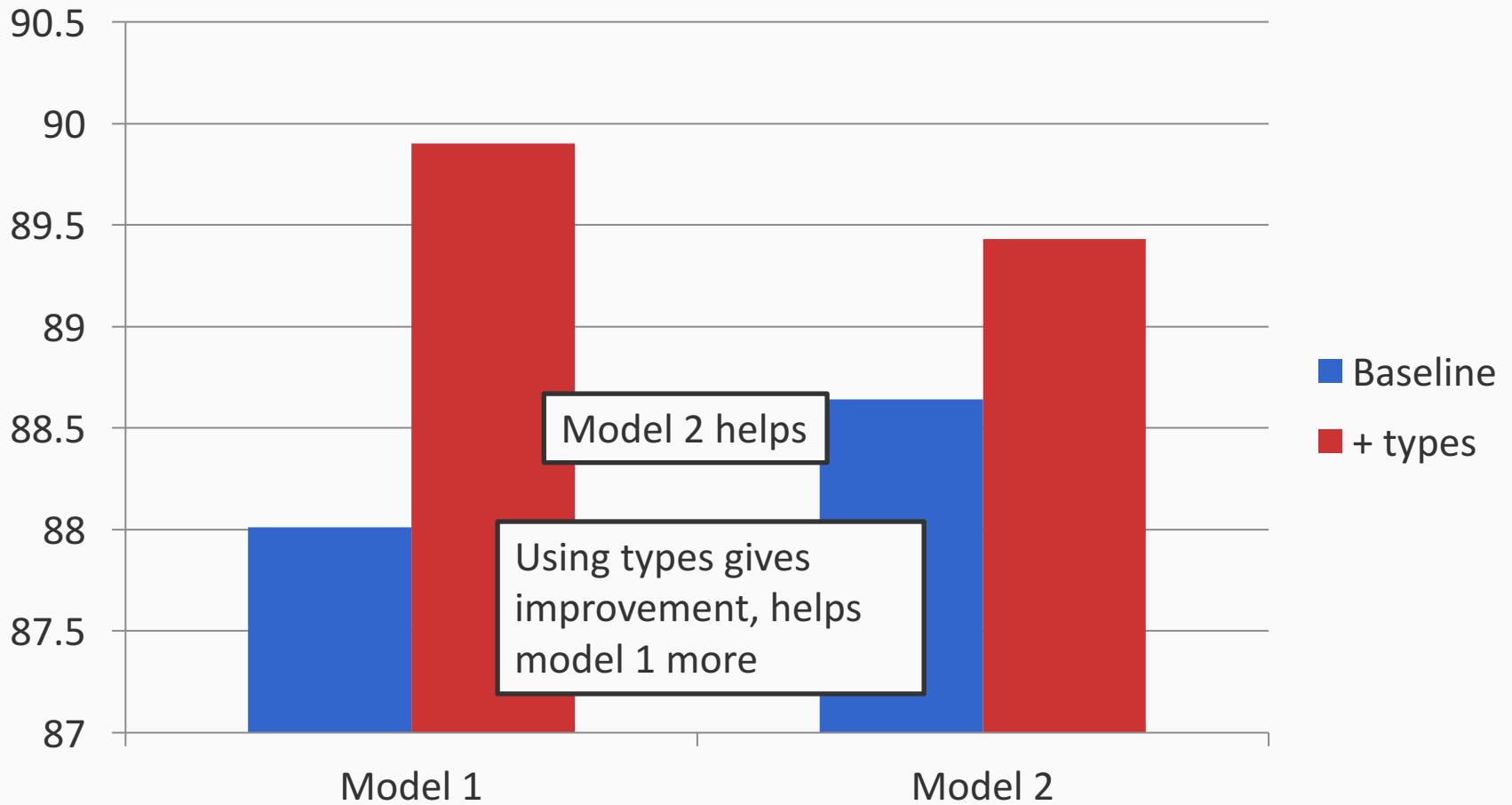
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Two models

- Model 1
 - Predict only relation label: Multi-class
 - Use features from all possible governor and object candidates
 - Also types
- Model 2 uses features from the structure
 - Predict full structure: relation and arguments
 - Also types

Accuracy of supersense labeling



What do we have?

Input	Relation	Governor type	Object type
Died of pneumonia	Cause	Experience	Disease
Suffering from flu	Cause	Experience	Disease
Recovered from flu	StartState	Change	Disease

Governor, object and their types as a certificate for the choice of relation label

Summary

- Prepositions express a diverse set of relations
 - An ontology of preposition supersenses
 - Can enrich existing PropBank/FrameNet representation
- Models for predicting preposition relations
 - Arguments and types help

This talk

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Jonathan
Berant



Heather
Chen



Chris
Manning

Modeling Biological Processes for Reading Comprehension

EMNLP 2014

What can we do with a semantic representation?

Reading comprehension is hard!

Water is split, providing a source of electrons and protons (hydrogen ions, H^+) and giving off O_2 as a by-product. Light absorbed by chlorophyll drives a transfer of the electrons and hydrogen ions from water to an acceptor called $NADP^+$.

What can the splitting of water lead to?

A: Light absorption

B: Transfer of ions

Reading comprehension is hard!

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Enable

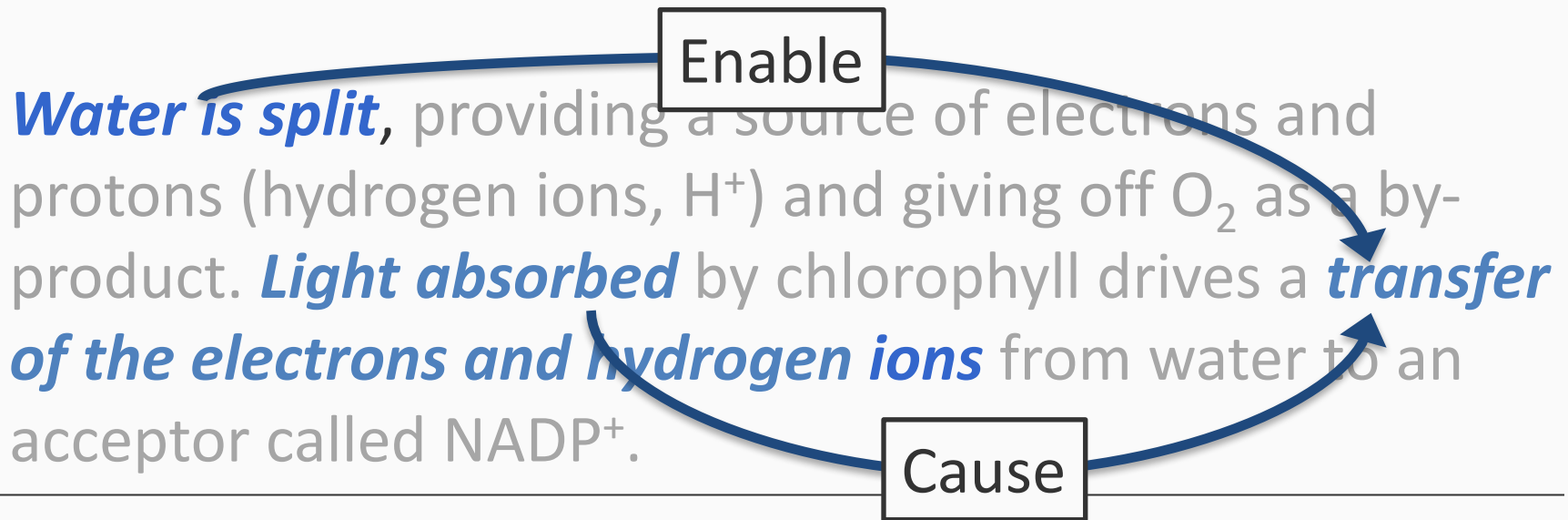
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The punchline

1. A new reading comprehension task requiring reasoning over processes

Processes are fundamental in many domains

2. A new dataset `ProcessBank` consisting of descriptions of biological processes with

- Rich process structure annotated, *and*
- Multiple-choice questions

3. A new end-to-end system for reading comprehension

- Predict structure and treat it as a knowledge base
- Parse question as query to this knowledge base (semantic parsing)

A new dataset: ProcessBank

Motivation: macro vs. micro reading

- Macro reading:

- Exploits web-scale redundancy

[Etzioni et al., 2006, Carlson et al., 2010, Fader et al., 2011]

- Factoid questions

[Berant et al., 2014, Fader et al., 2014]

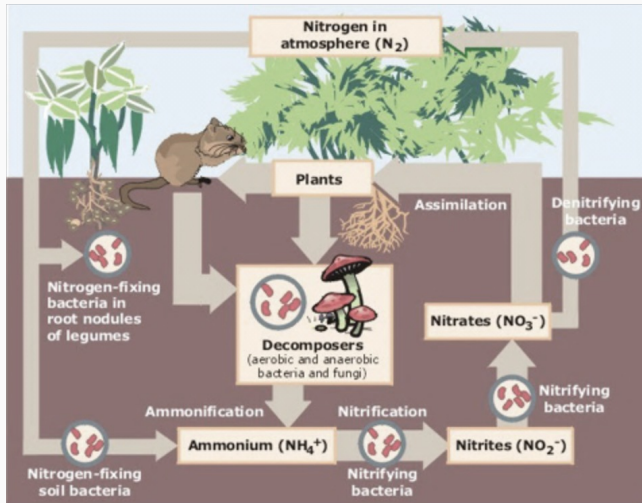
- Micro reading:

- Single document
- Requires reasoning
- Non-factoid questions

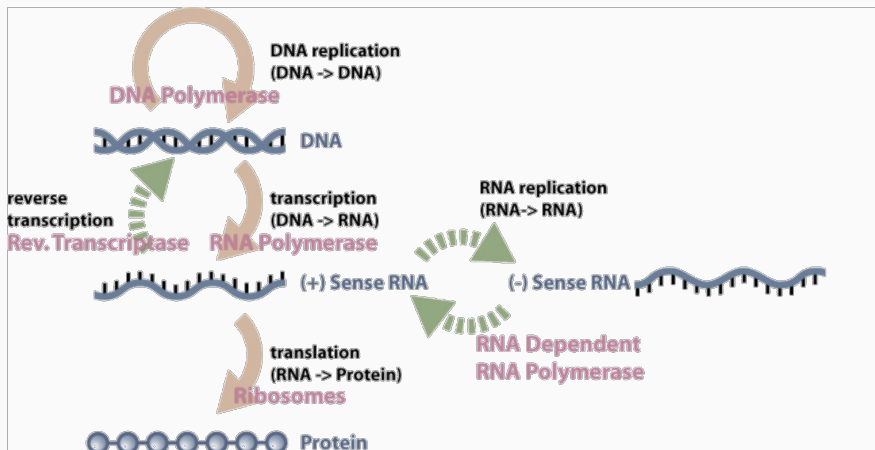
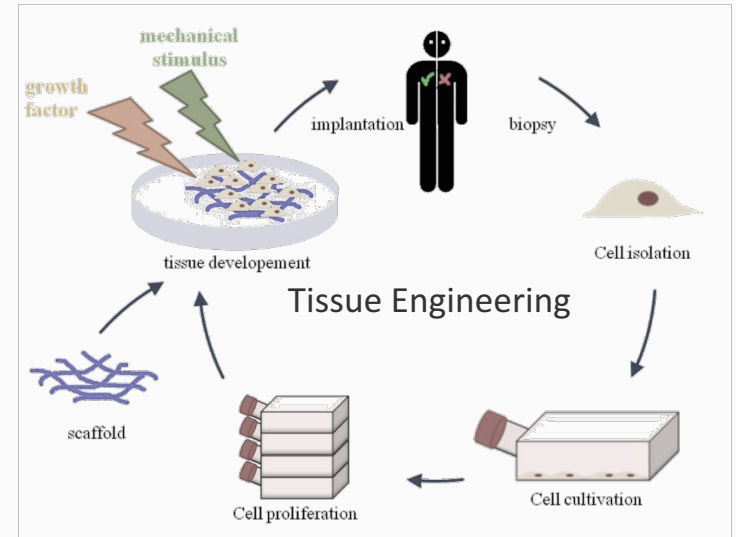
[Richardson et al., 2013, Kushman et al., 2014]

Chosen domain:
**Biological process
descriptions**

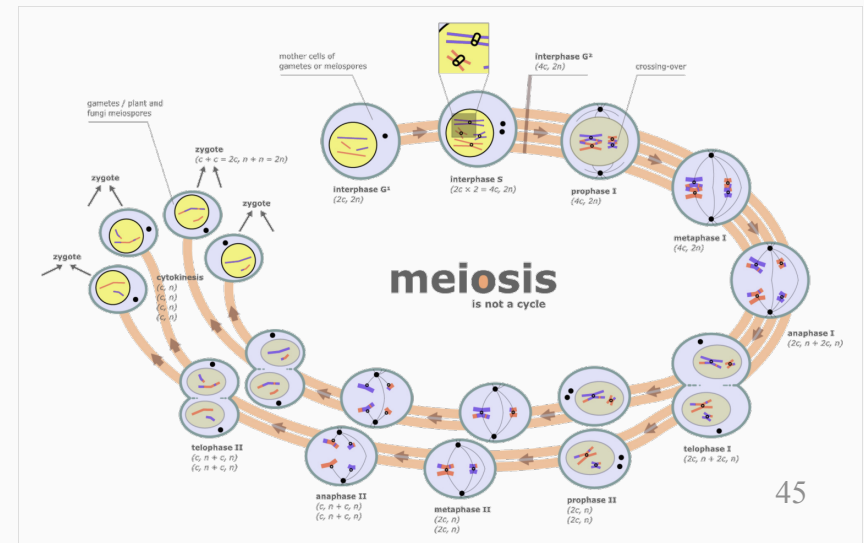
Processes abound in biology



Nitrogen Cycle



Central Dogma of Molecular Biology



Creating a difficult reading comprehension task

200 paragraphs from the textbook *Biology*

[Campbell & Reese, 2005]

Desiderata

1. Test understanding of inter-relations between events and entities
2. Both answers should have similar lexical overlap:
 - Trump shallow approaches
 - Sidestep lexical variability

Reading comprehension annotation

- Annotation instructions: Ask questions about events, entities and their relationships
 - 10 examples provided
 - Two answer choices, only one unambiguously correct
- 200 paragraphs → 585 questions
- Second annotator answered the questions
 - 98.1% agreement

Examples of annotated questions

Dependencies between events/entities (70%)

Q: *What can the splitting of water lead to?*

A: Light absorption

B: Transfer of ions

Temporal ordering of events (10%)

Q: *What is the correct order of events?*

A: PDGF binds to tyrosine kinases, then cells divide, then wound healing

B: Cells divide, then PDGF binds to tyrosine kinases, then wound healing

True-False questions (20%)

Q: *Cdk associates with MPF to become cyclin*

A: True

B: False

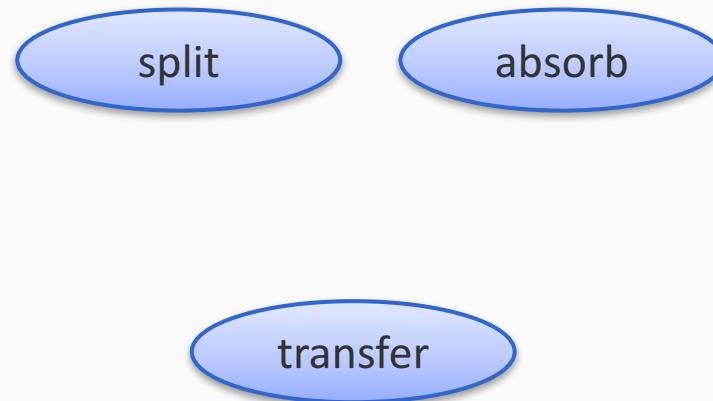
A second layer of annotation:

Process structures

Water is split, providing a source of electrons and protons (hydrogen ions, H^+) and giving off O_2 as a by-product. *Light absorbed* by chlorophyll drives a *transfer of the electrons and hydrogen ions* from water to an acceptor called $NADP^+$.

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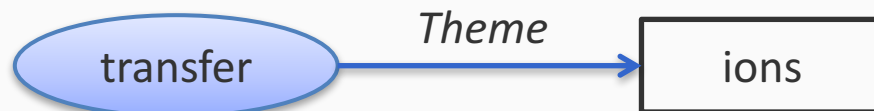
Triggers: Tokens
denoting occurrence
of an event

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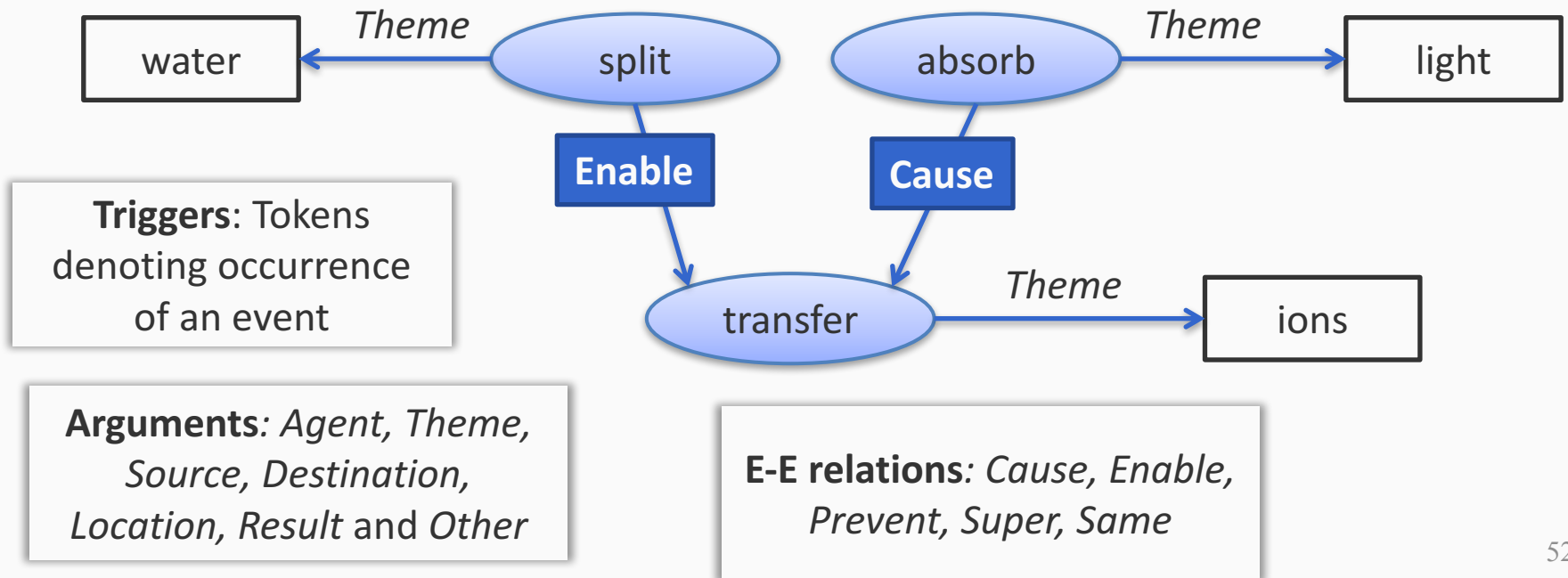
Triggers: Tokens denoting occurrence of an event



Arguments: Agent, Theme, Source, Destination, Location, Result and Other

A second layer of annotation: Process structures

Water is split, providing a source of electrons and protons (hydrogen ions, H^+) and giving off O_2 as a by-product. *Light absorbed* by chlorophyll drives a *transfer of the electrons and hydrogen ions* from water to an acceptor called $NADP^+$.



Process structure data

- Same 200 paragraphs from *Biology*
 - Paragraphs annotated and verified
- Three annotators
 - Biologists
 - Independent from QA annotator
 - Potentially conflicting with questions
- More nuances
 - Eg: No temporal ordering of events

What is ProcessBank?

- 200 paragraphs from the textbook *Biology*
 - Manually chosen to represent biological processes
- Each paragraph annotated with
 1. Non-factoid reading comprehension questions
 2. Process structures

Answering questions: Overview

System in a nutshell

Water is split, providing a source of electrons and protons (hydrogen ions, H^+) and giving off O_2 as a by-product. *Light absorbed* by chlorophyll drives a *transfer of the electrons and hydrogen ions* from water to an acceptor called $NADP^+$.

What can the splitting of water lead to?

A: Light absorption

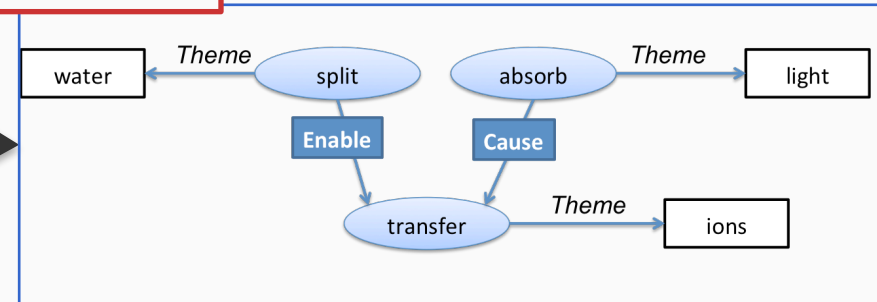
B: Transfer of ions

System in a nutshell

Process Structure Prediction

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Step 1



What can the splitting of water lead to?

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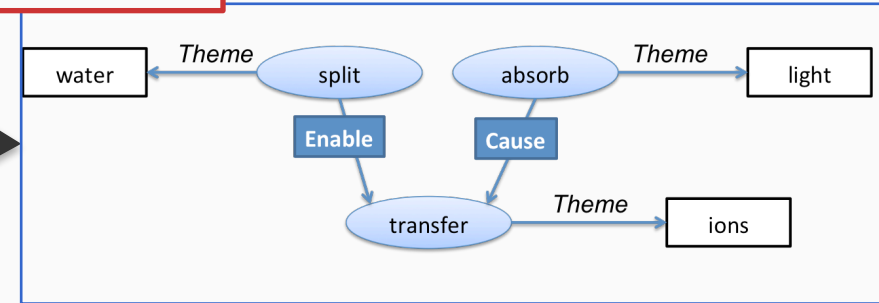
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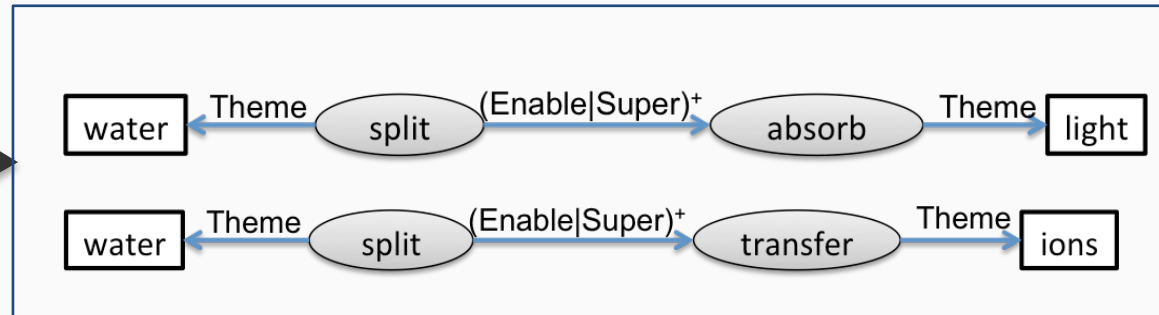
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Step 2

Question Parsing

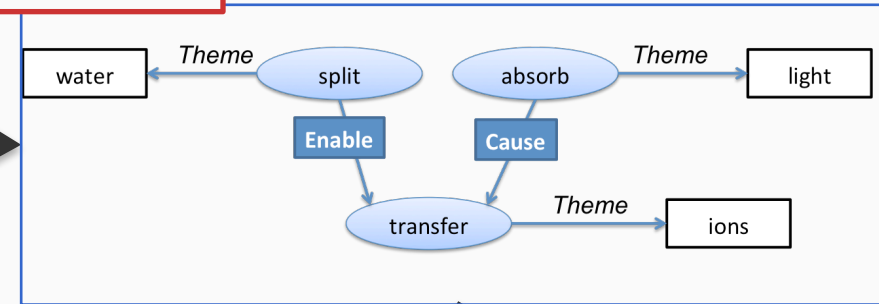


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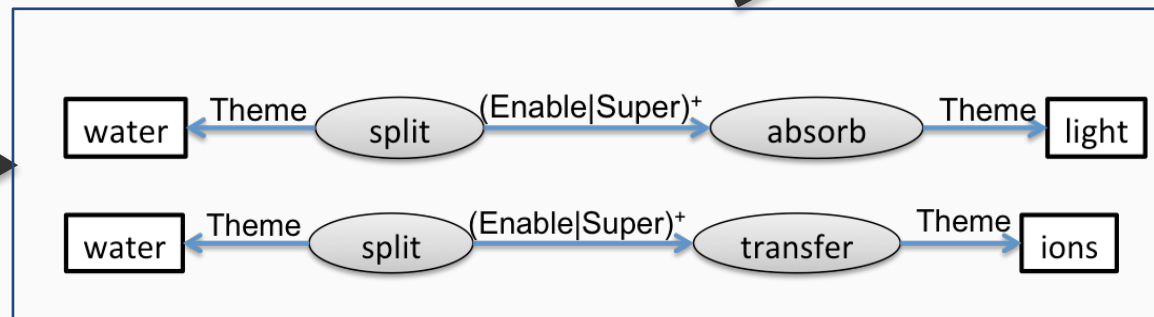
B: Transfer of ions

Answering Question

Step 3: Answer = B

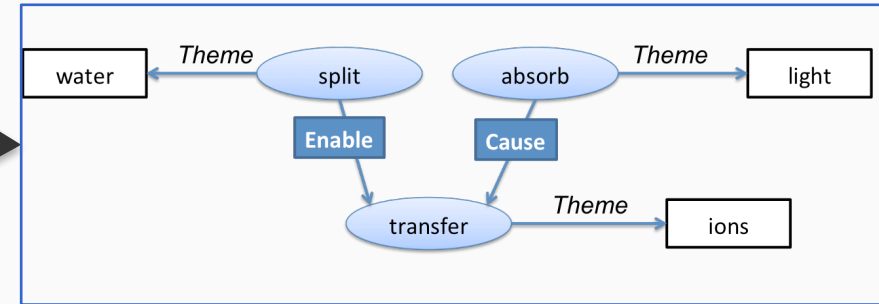
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Step 1



Predicting process structures

Process structure prediction

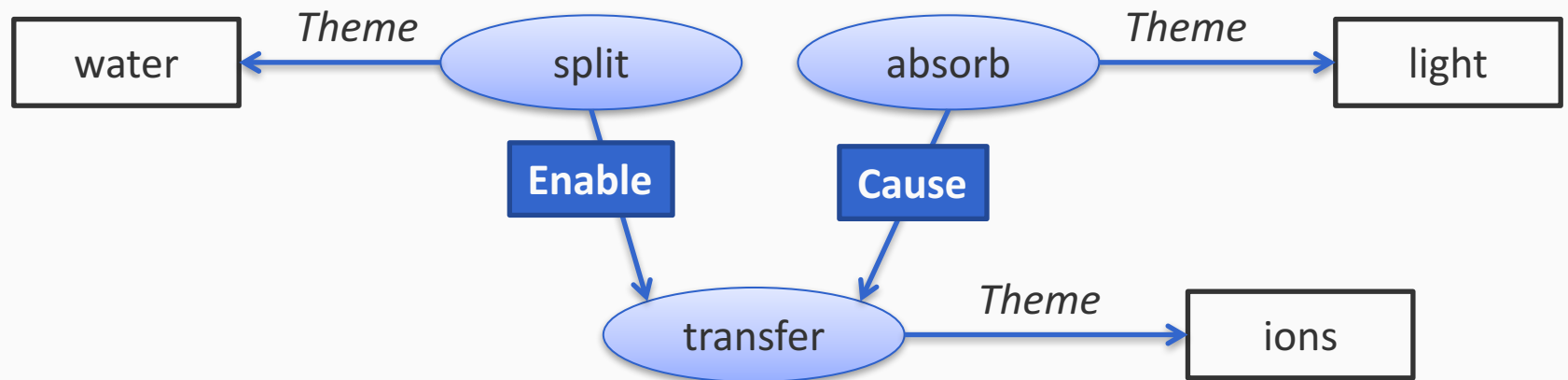
1. Train trigger identifier

Logistic regression; features from words, lists

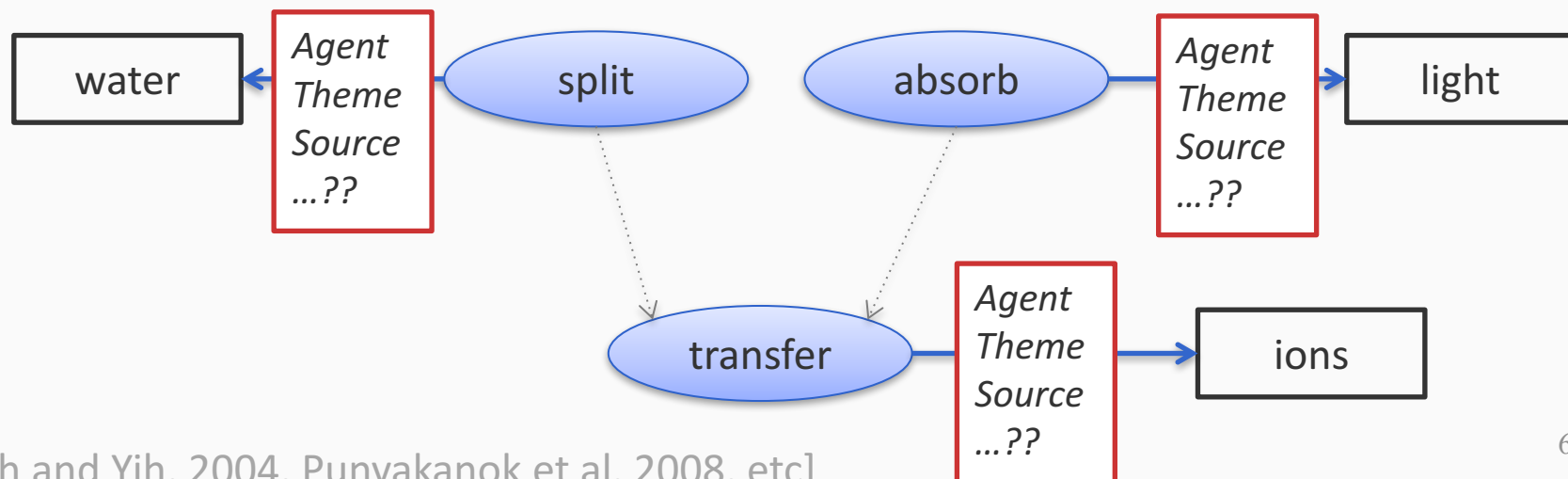
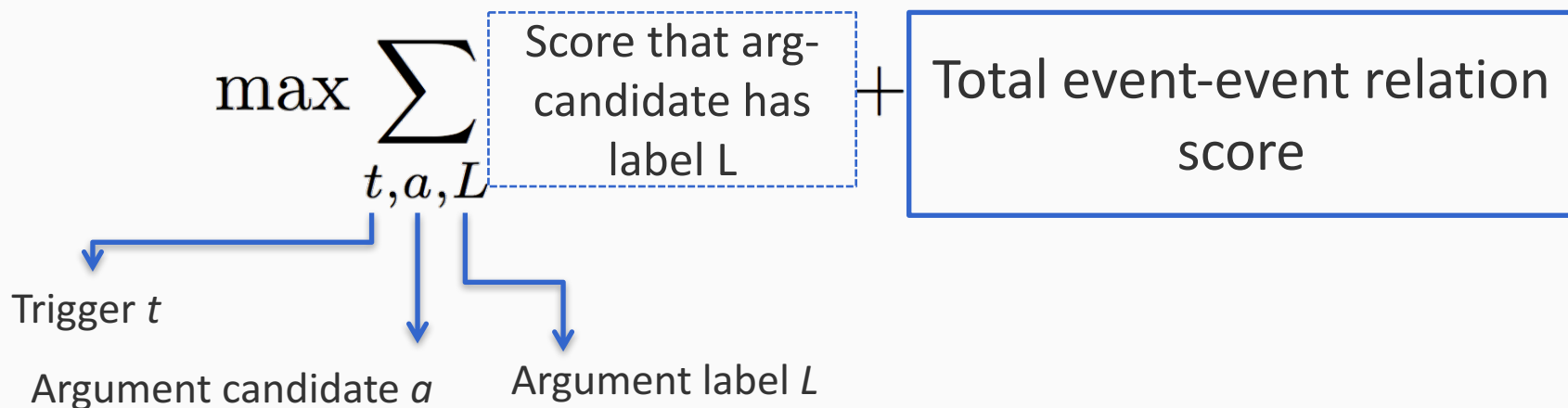
2. Joint learning and inference over arguments and event-event relations using predicted triggers

Event-arguments and event-event relations

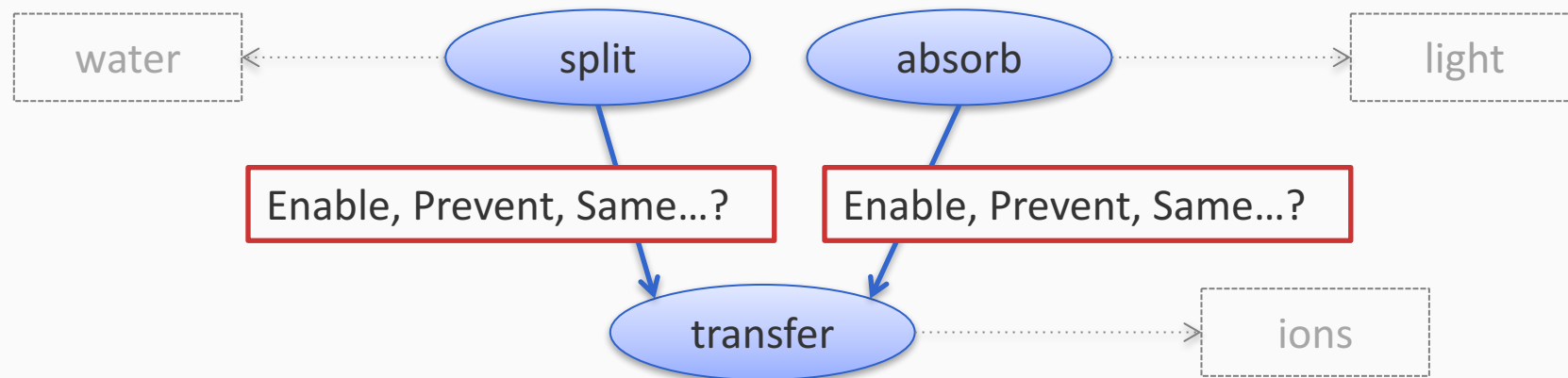
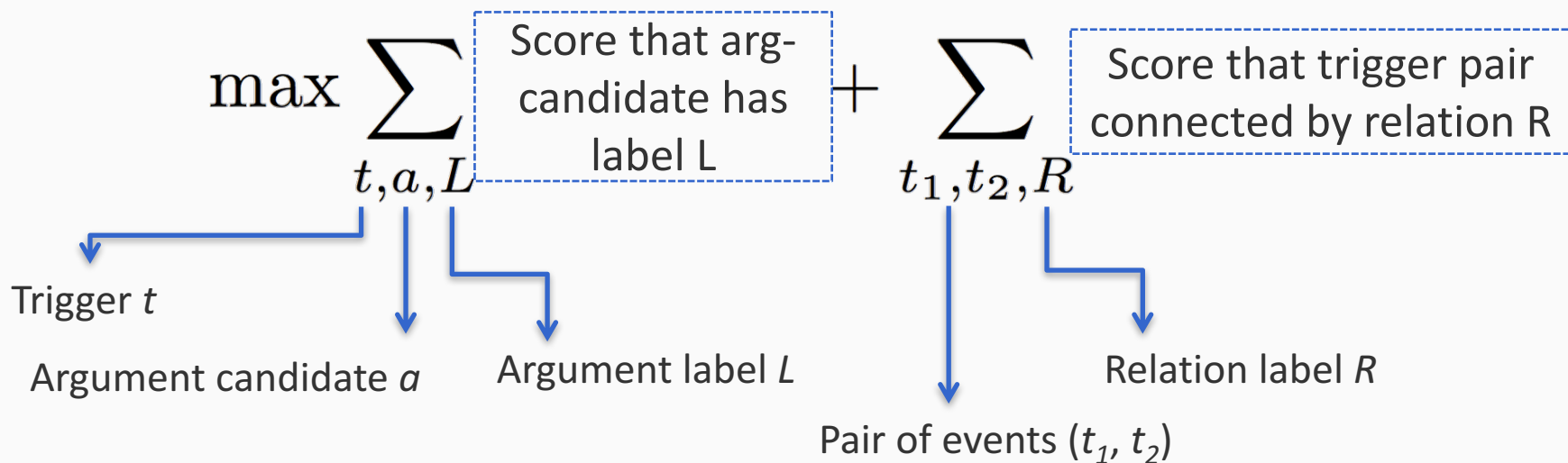
$$\max \left[\begin{array}{|c|} \hline \text{Total event-argument score} \\ \hline \end{array} + \begin{array}{|c|} \hline \text{Total event-event relation score} \\ \hline \end{array} \right]$$



Event-arguments and event-event relations



Event-arguments and event-event relations



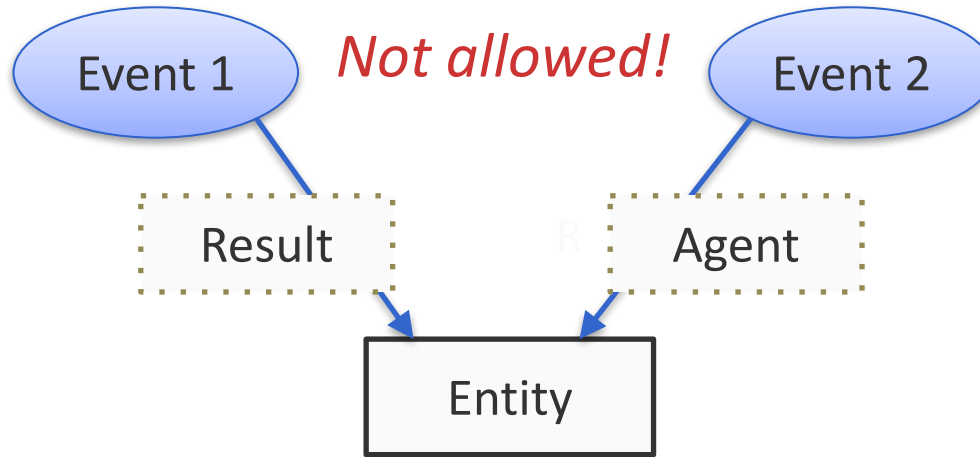
Joint inference with constraints

1. No overlapping arguments
2. Maximum number of arguments per trigger
3. Maximum number of triggers per entity
4. Connectivity
5. Events that share arguments must be related

And a few other constraints

Joint inference with constraints

1.
2.
3.
4.

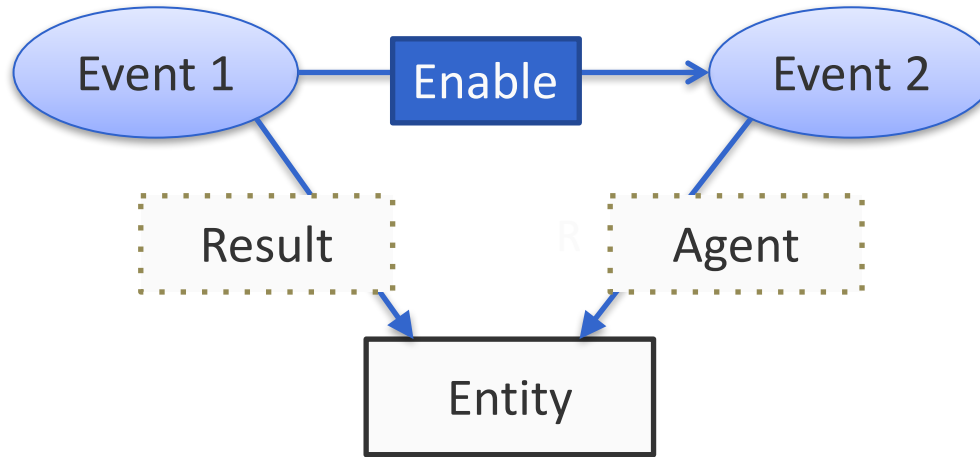


5. *Events that share arguments must be related*

And a few other constraints

Joint inference with constraints

1.
2.
3.
4.



5. *Events that share arguments must be related*

And a few other constraints

Learning and Inference

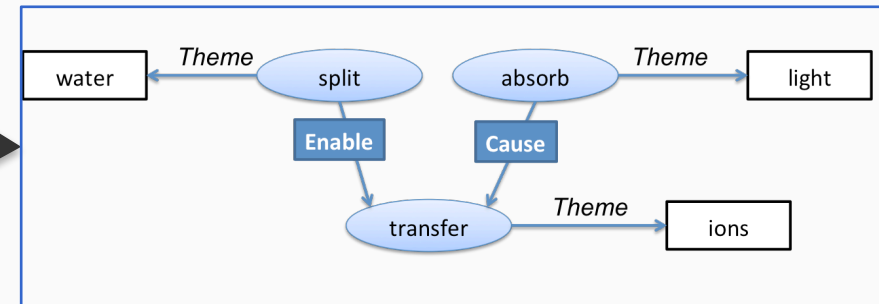
- Linear model to score argument labels and event-event relations
 - Related: Semantic role labeling, information extraction
- Structured averaged perceptron
- Gurobi ILP solver (exact solution)

Answering questions

Where are we?

Water is split, providing a source of electrons and protons (hydrogen ions, H^+) and giving off O_2 as a by-product. *Light absorbed* by chlorophyll drives a *transfer of the electrons and hydrogen ions* from water to an acceptor called $NADP^+$.

Step 1



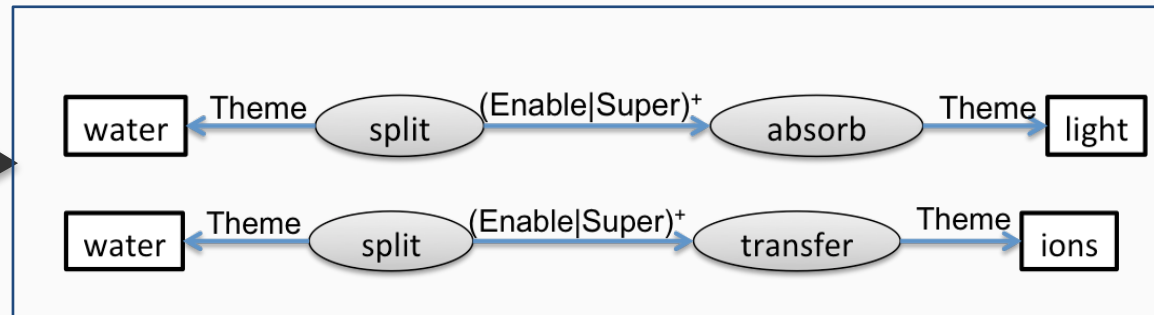
What can the splitting of water lead to?

A: Light absorption

B: Transfer of ions

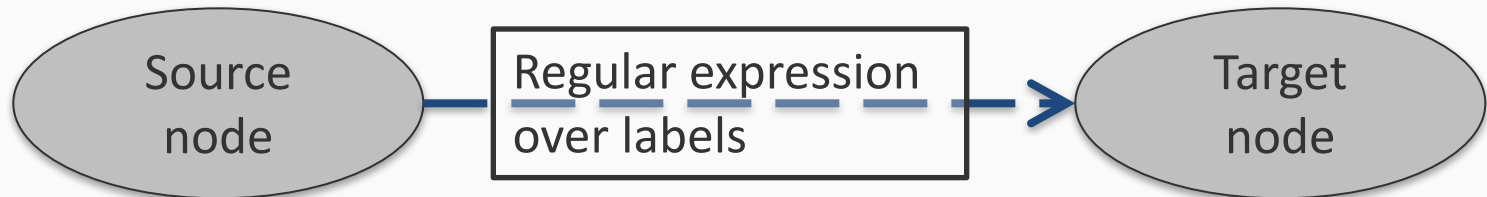
Step 2

Question Parsing



Question parsing

- Task: Given a question and two answers, produce two queries
 - One for each answer
- Query structure

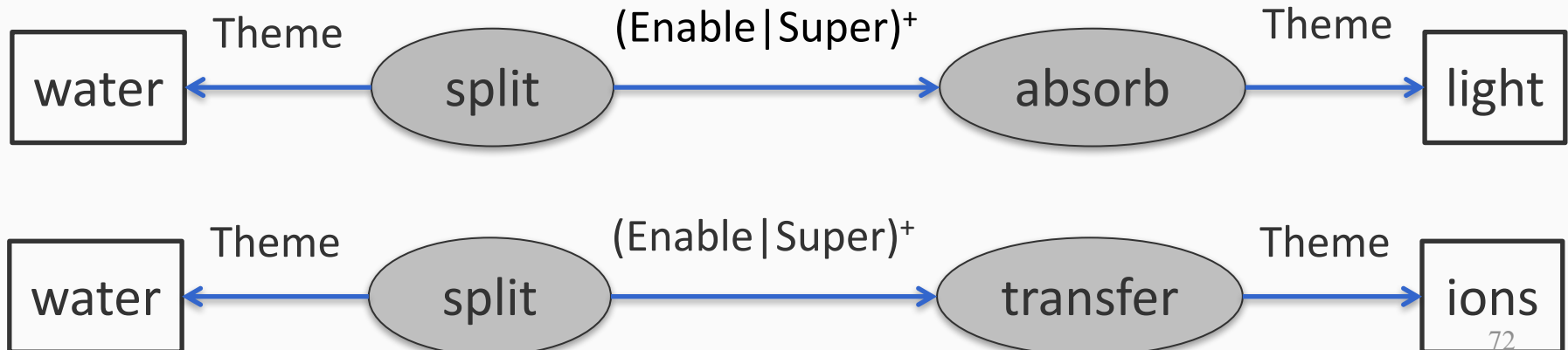


Parsing question to produce formal queries

What does the splitting of water lead to?

A: Light absorption

B: Transfer of ions



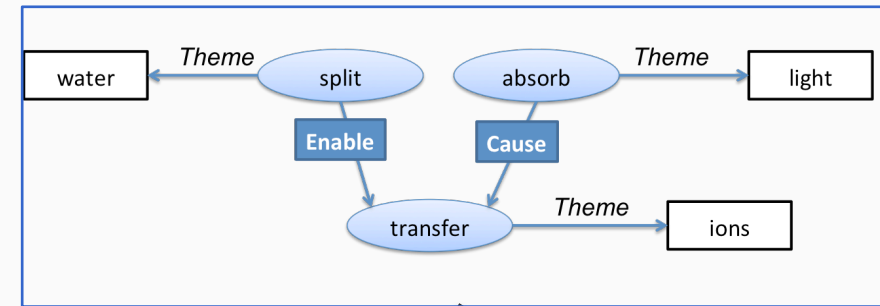
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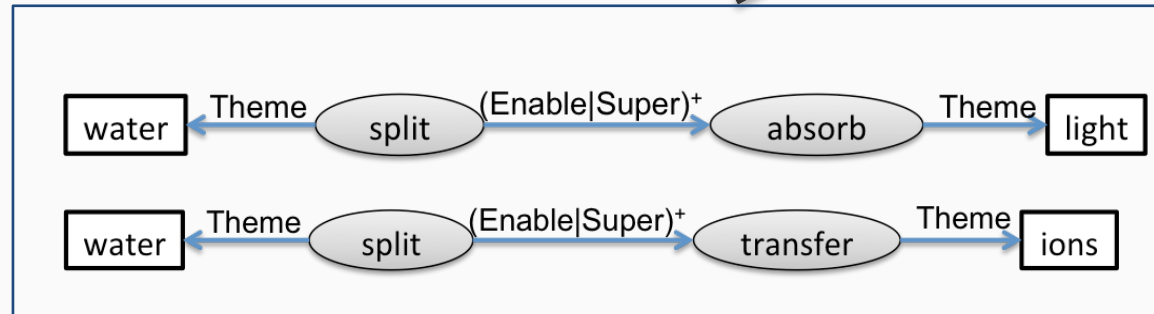
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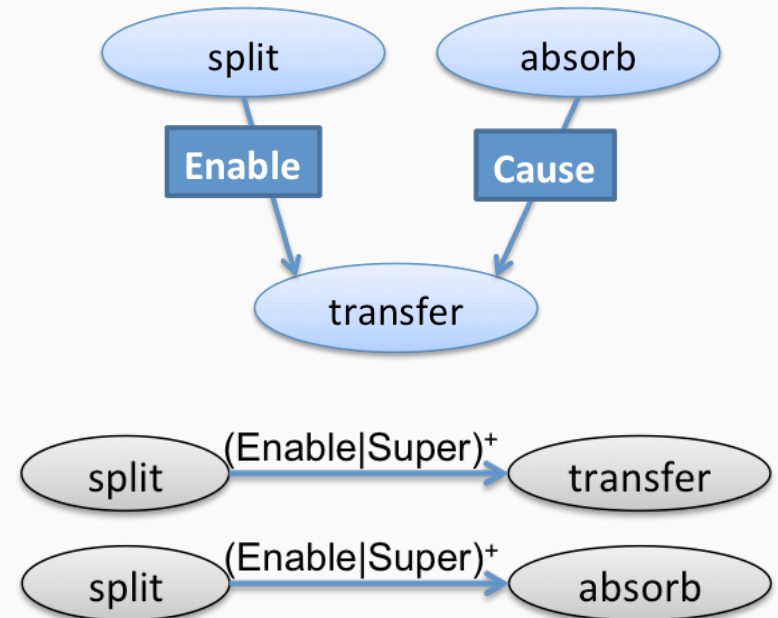
Answering Question

Step 3: Answer = B



Step 3: Answering questions

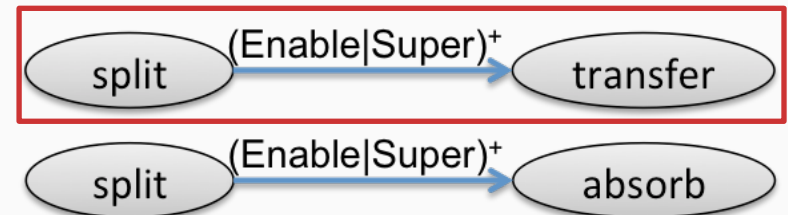
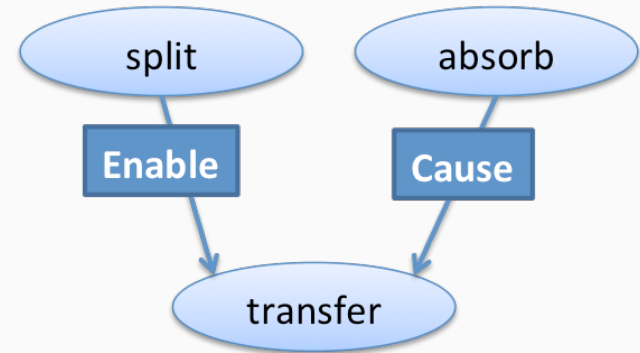
- Given
 - Process structure
 - Two queries



- Answering algorithm

Step 3: Answering questions

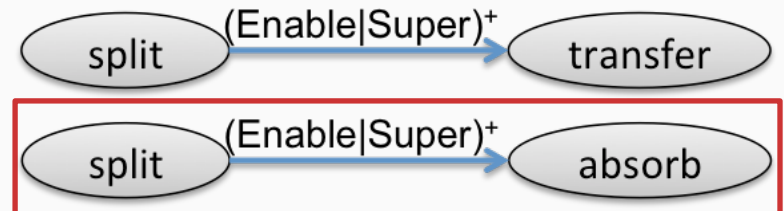
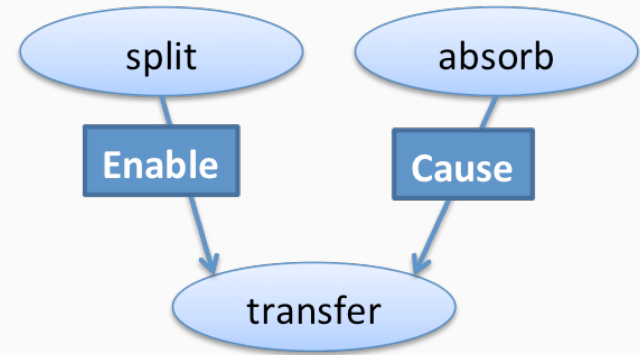
- Given
 - Process structure
 - Two queries



- Answering algorithm
 1. Find matching path (valid proof)

Step 3: Answering questions

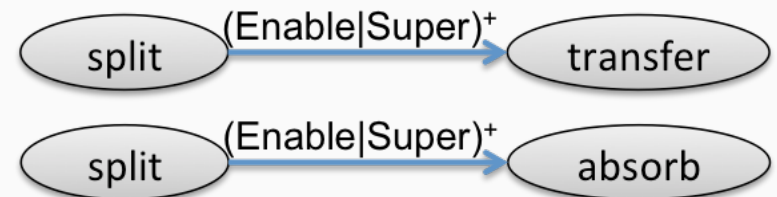
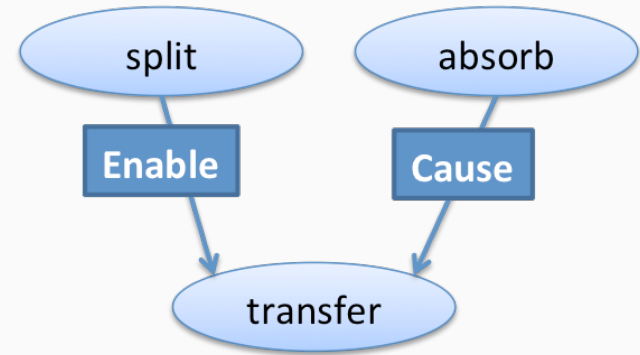
- Given
 - Process structure
 - Two queries



- Answering algorithm
 1. Find matching path (valid proof)
 2. Else, find contradiction of causality (refutation)

Step 3: Answering questions

- Given
 - Process structure
 - Two queries



- Answering algorithm
 1. Find matching path (valid proof)
 2. Else, find contradiction of causality (refutation)
 3. Back off to baseline

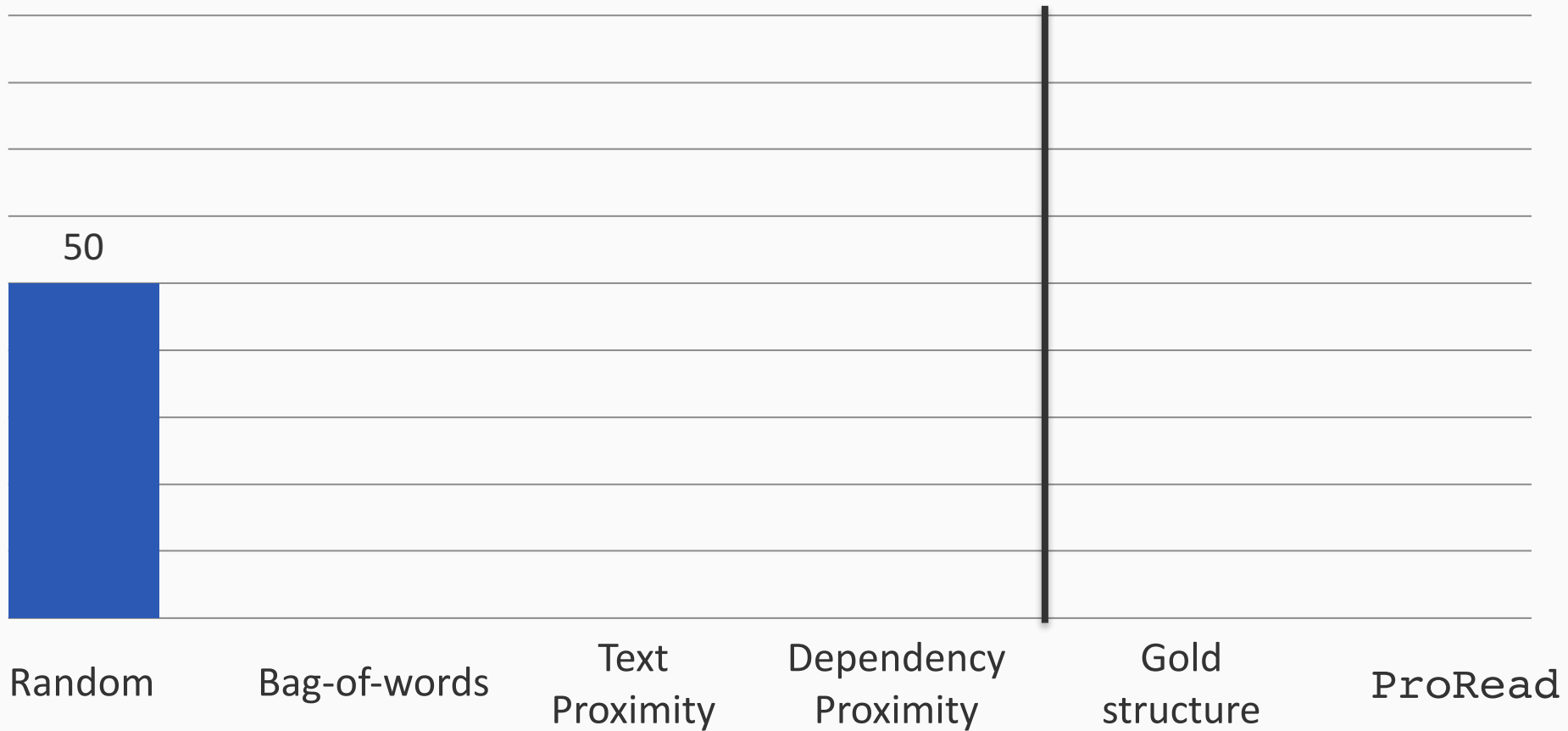
Experiments and Results

Question Answering Accuracy

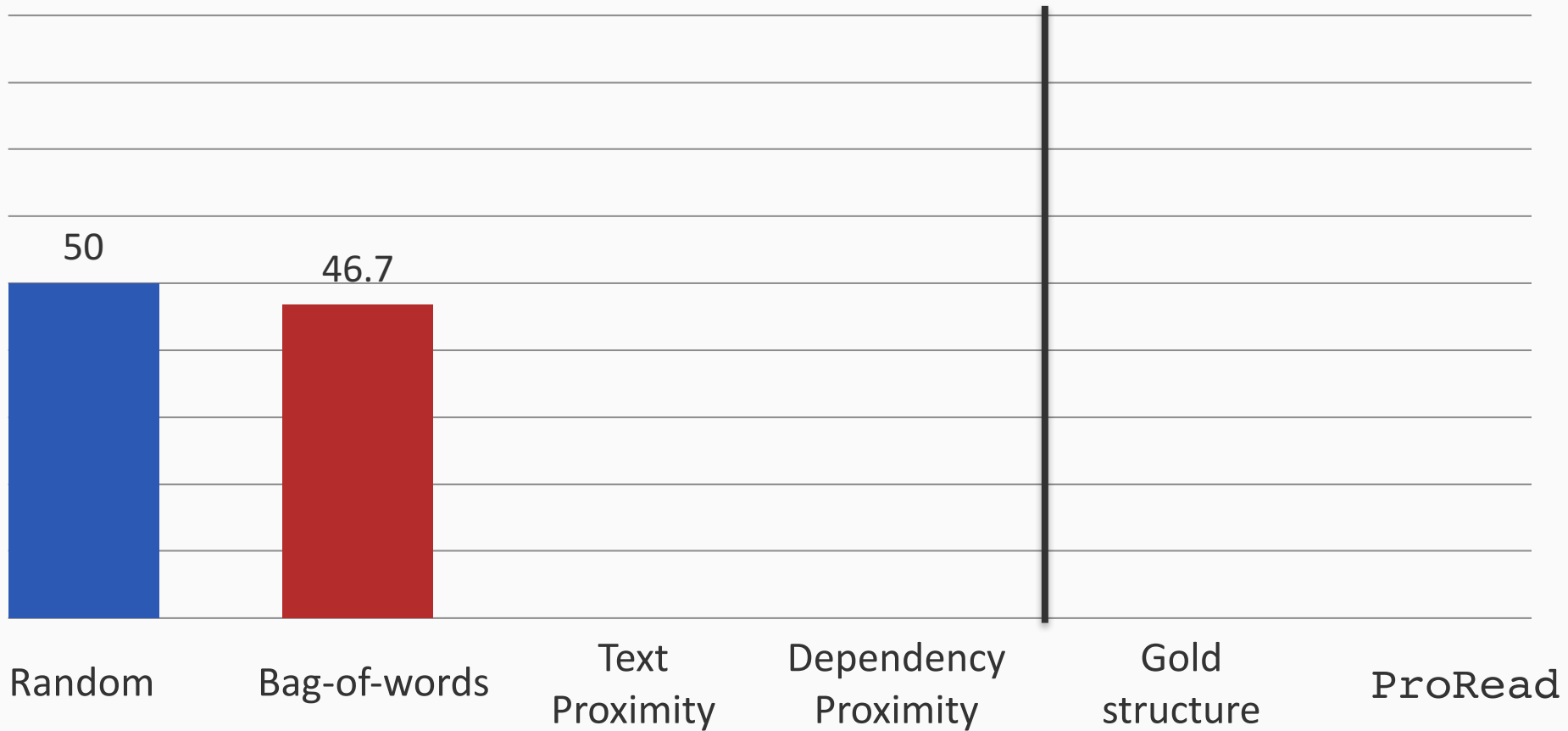
Train: 150 processes
Test: 50 processes

Random Bag-of-words Text Proximity Dependency Proximity Gold structure ProRead

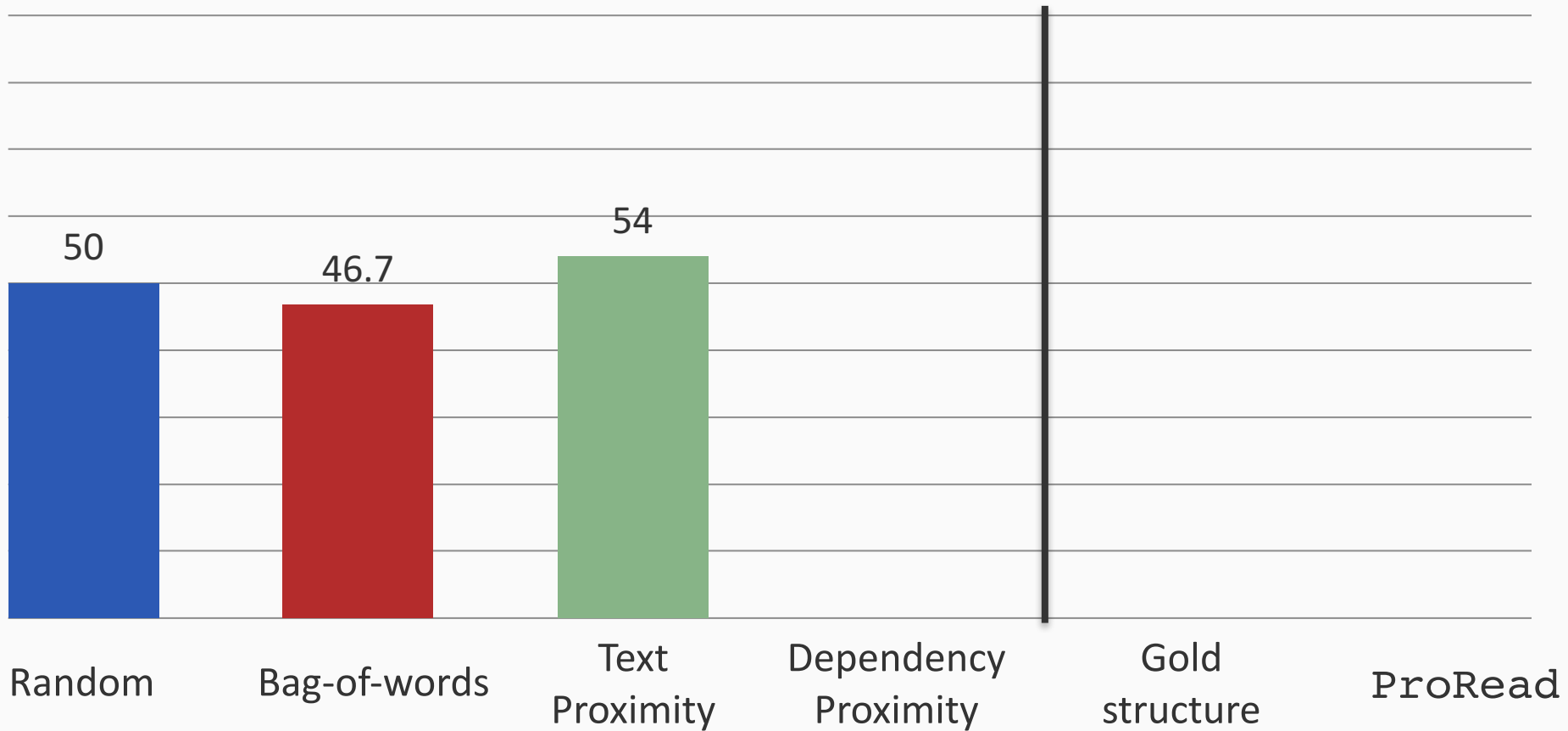
Question Answering Accuracy



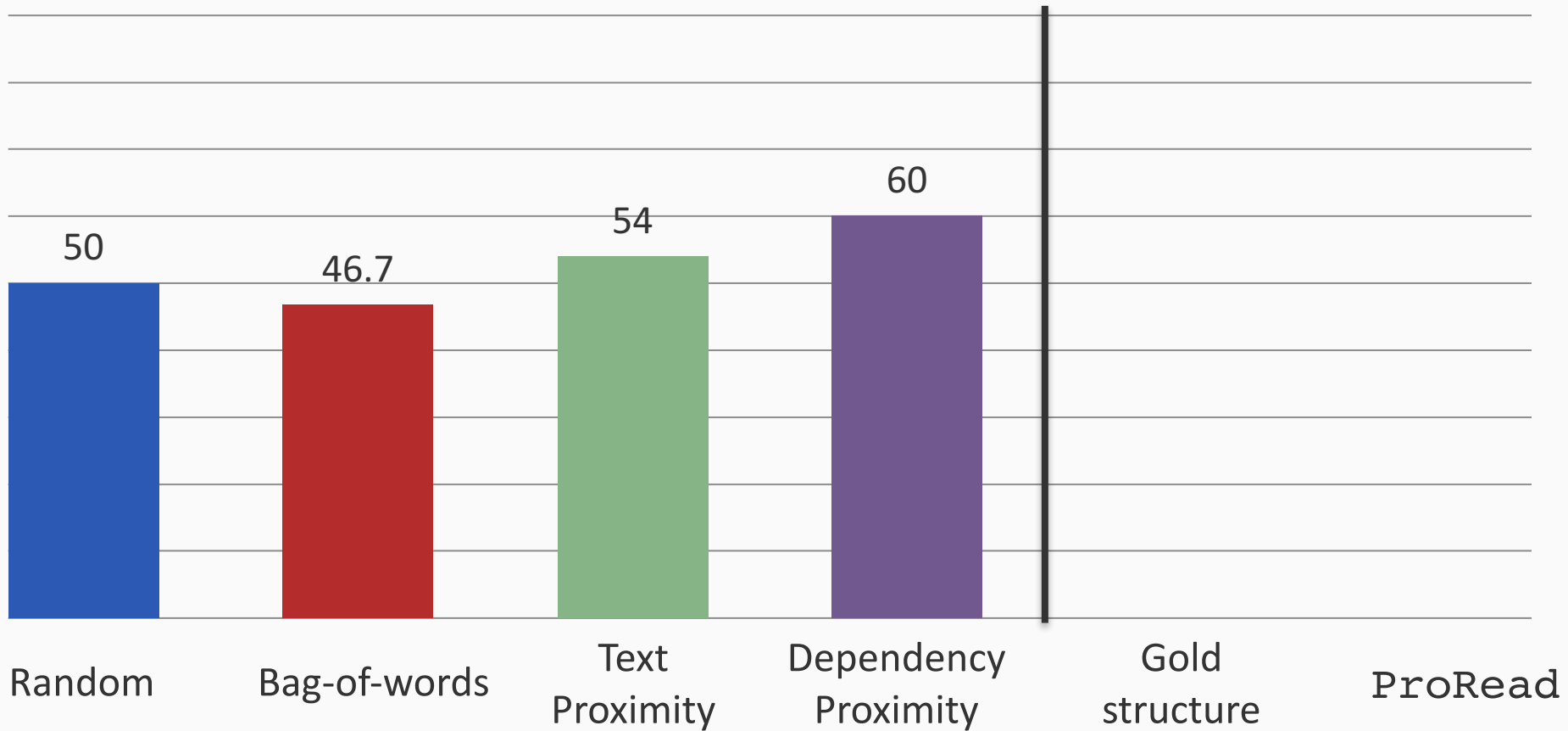
Question Answering Accuracy



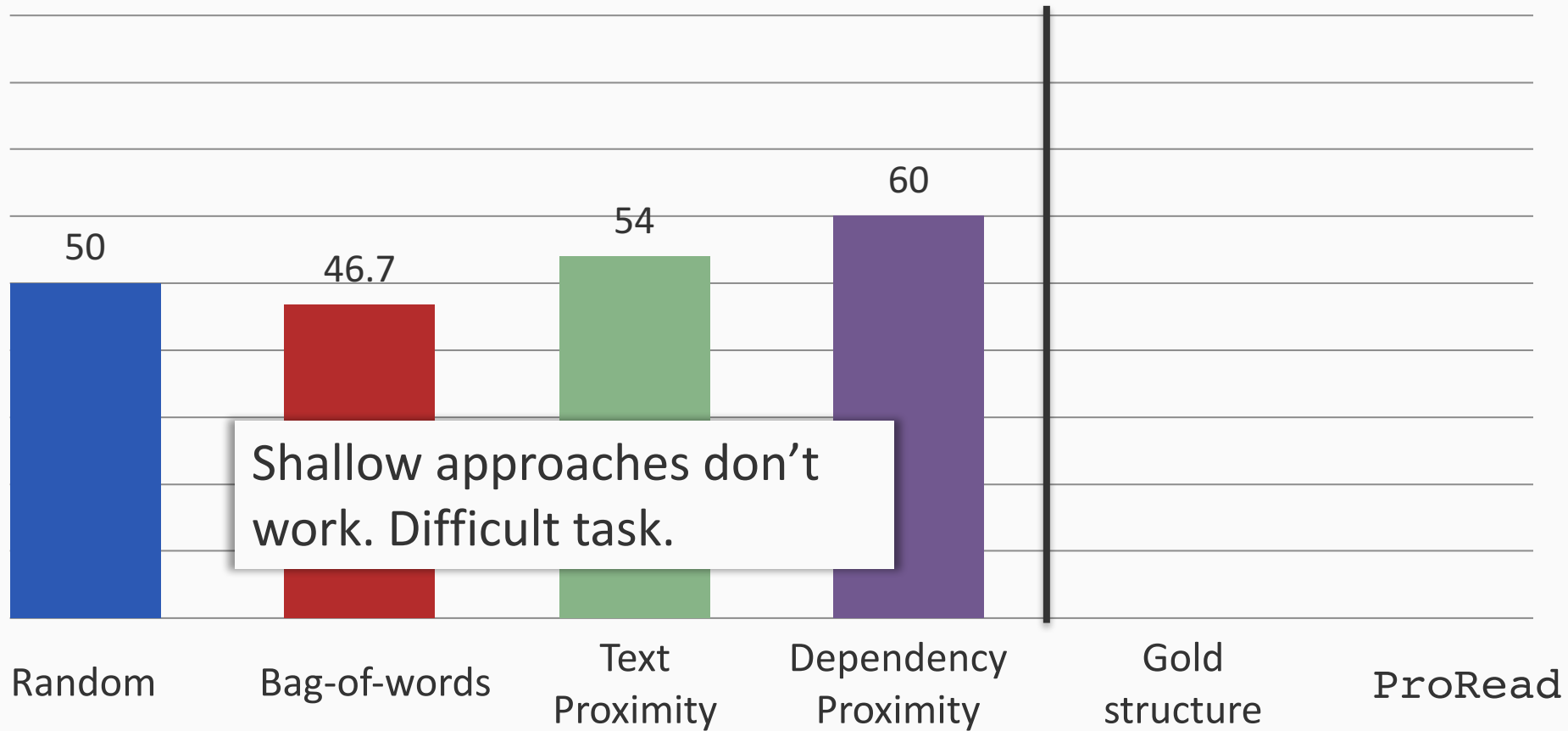
Question Answering Accuracy



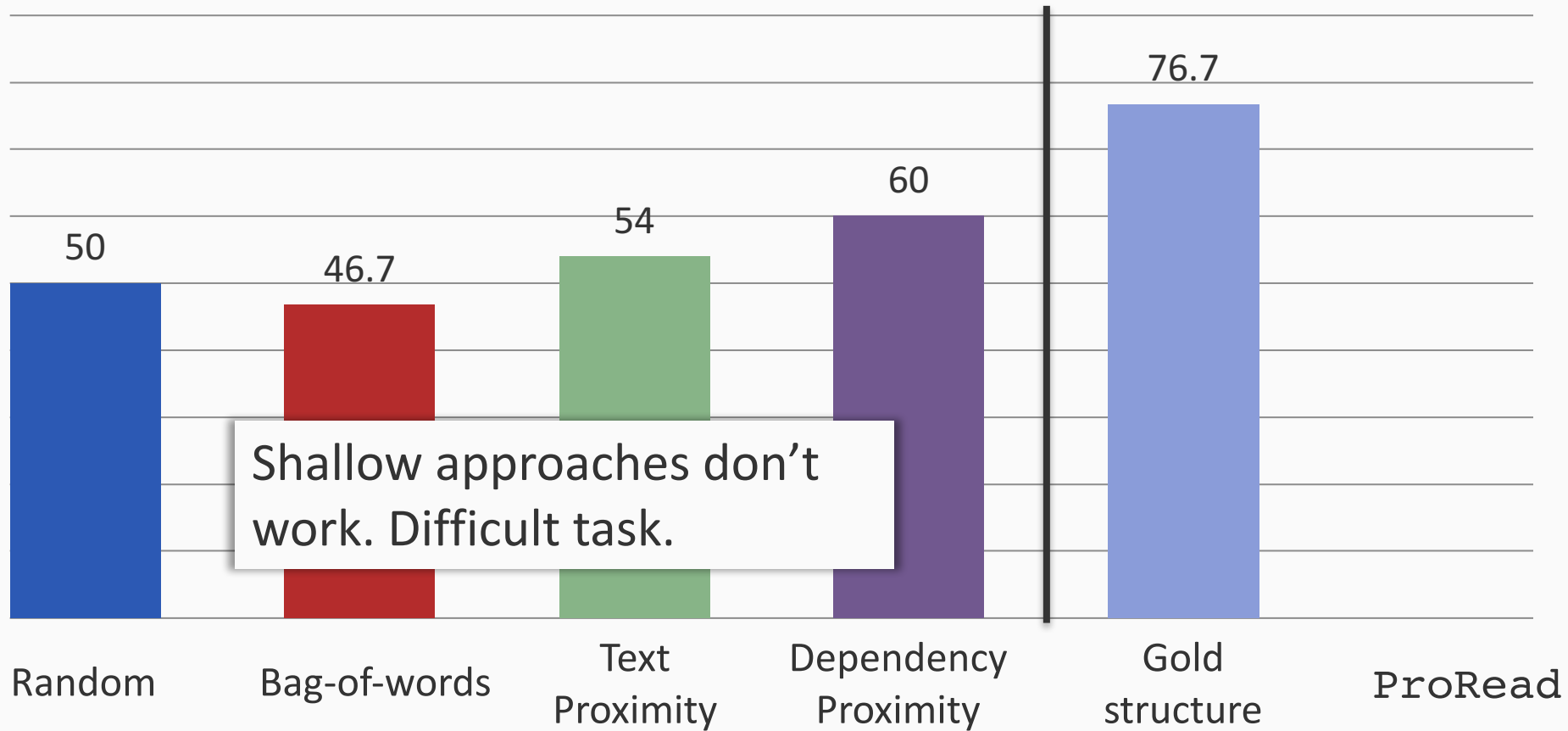
Question Answering Accuracy



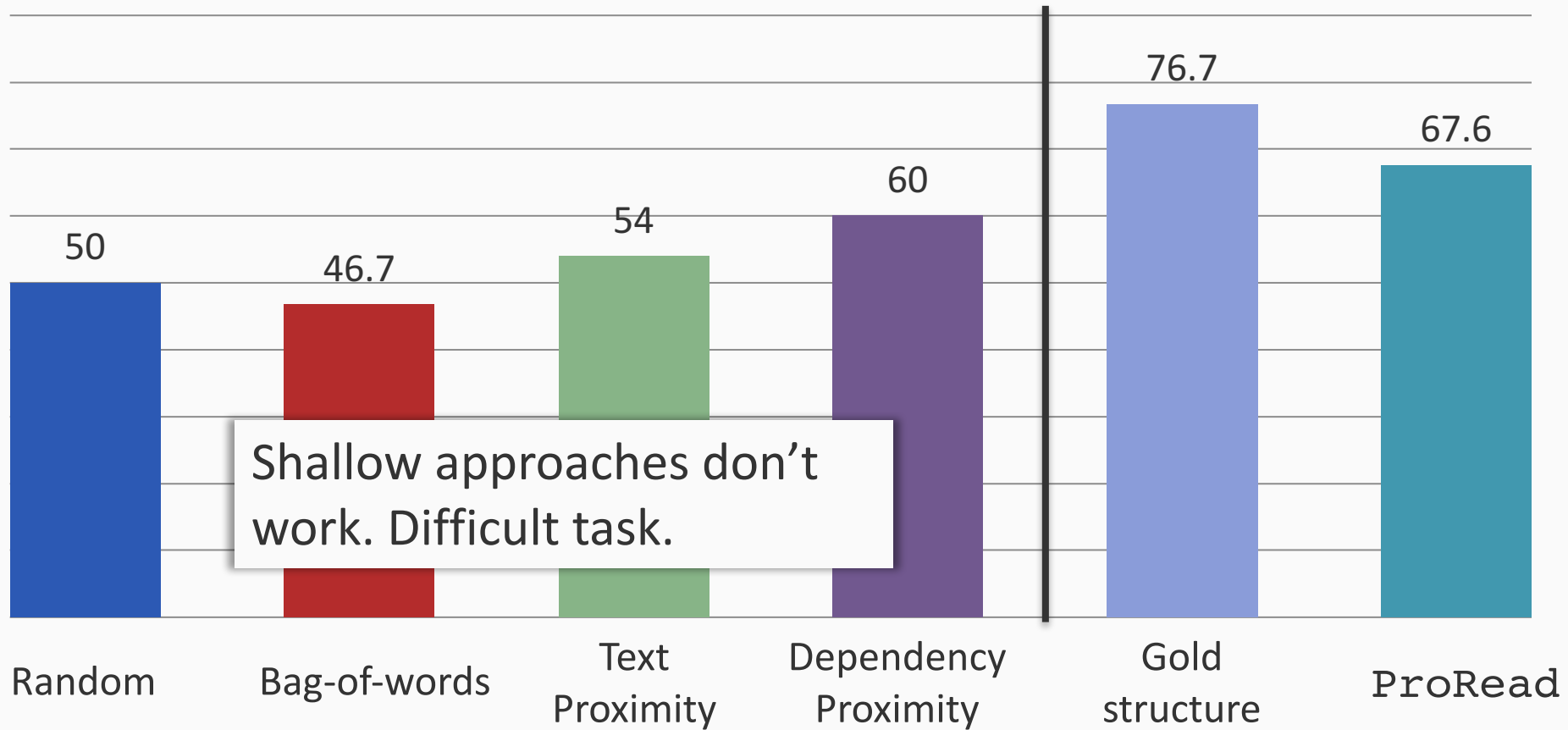
Question Answering Accuracy



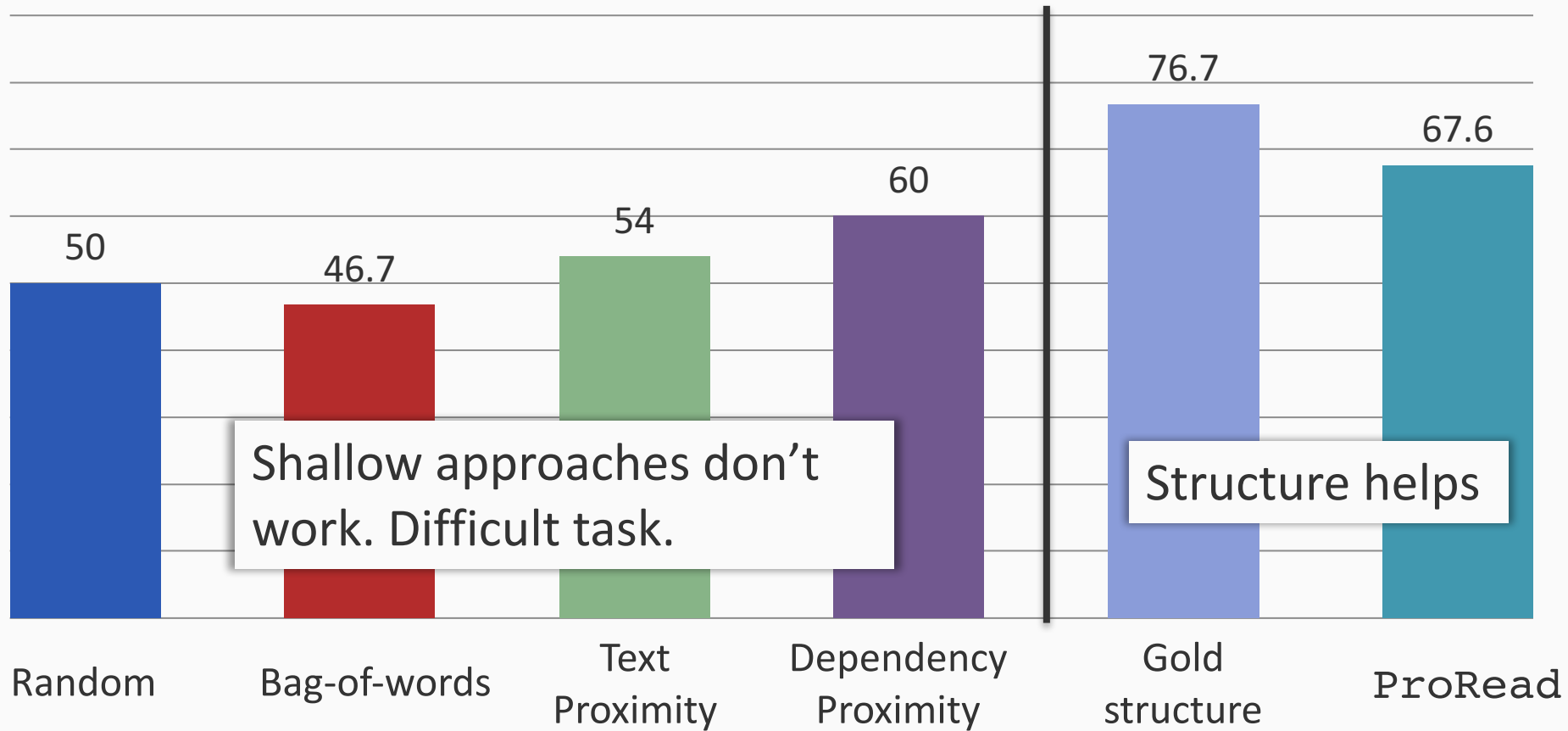
Question Answering Accuracy



Question Answering Accuracy



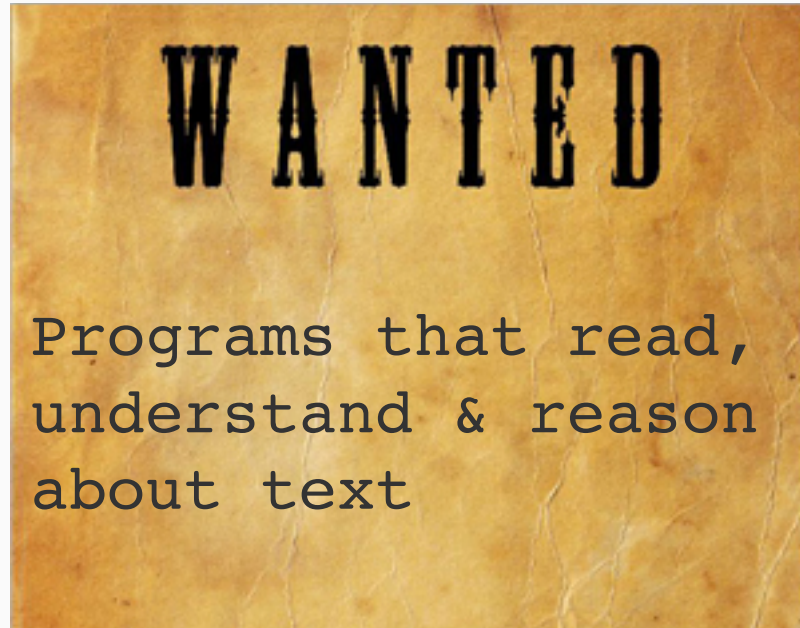
Question Answering Accuracy



Summary

- A new reading comprehension task
- A dataset of structures with Q&A: ProcessBank
- An end-to-end system for question answering via predicted structures
- Rich entity and event structure helps

Bigger picture



Need linguistically motivated structured representations for machine reading