

Studying the anticipation of QUDs and discourse relations by crowdsourcing a dataset of 'evoked questions'

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Why wouldn't they use their prostheses?

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Why do they not use their prostheses?

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Why didn't amputees use their prostheses?

Research questions

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Questions Under Discussion

e.g. Roberts 2012

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- Discourse is (can be modeled as) a process of raising and resolving questions. *≠interrogatives*
- These questions can be implicit or made (partially) explicit.
- The QUD of a given discourse move is the question it targets.
- Discourse progresses from question to question in a reasonable/rational/natural way.

... and evoked questions?



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- If part of a discourse strongly *evokes* a certain question...
- ...and this question happens to be answered by the next discourse move...



... and evoked questions?



- If part of a discourse strongly *evokes* a certain question...
- ...and this question happens to be answered by the next discourse move...
- ...then that question is very likely its QUD.

The example again

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Why do they not use their prostheses?

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Why wouldn't they use their prosthesis?

How old are you now?

Why didn't amputees use their prostheses?

The reason, I would come to find out, was that their prosthetic sockets were painful [...]

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Unless we do it via evoked questions!

Elicitation tool

spellout.net/ibexexps/mwestera/evoque/

Elicitation tool

Person A:

Och the wee mite

Person B:

I know she's uh

And also as well I'm getting really worried

Everybody keeps going on **how wee she is**

► **Please enter a question the text evokes for you at this point.**

(The text so far must *not* yet contain an answer to the question!)

Is she really small?

► **In the text, **highlight** the main word or short phrase that evokes this question.**

Elicitation tool

Why is that hard? Well to see, let's imagine we take the Hubble Space Telescope and we turn it around and we move it out to the orbit of Mars. We'll see something like that, a slightly blurry picture of the Earth, because we're a fairly small telescope out at the orbit of Mars. Now let's move ten times further away. Here we are at the orbit of Uranus. It's gotten smaller, it's got less detail, less resolve. We can still see the little moon, but let's go ten times further away again.

Earlier, you also entered the following question:

How can the picture be improved?

► **Was that question answered in the new piece of text?**

Not answered at all. 1 2 3 4 5 *Completely answered.*

► **Enter the (complete/partial) answer in your own words:**

► **In the new piece of text, highlight the main word or short phrase suggesting this answer.**

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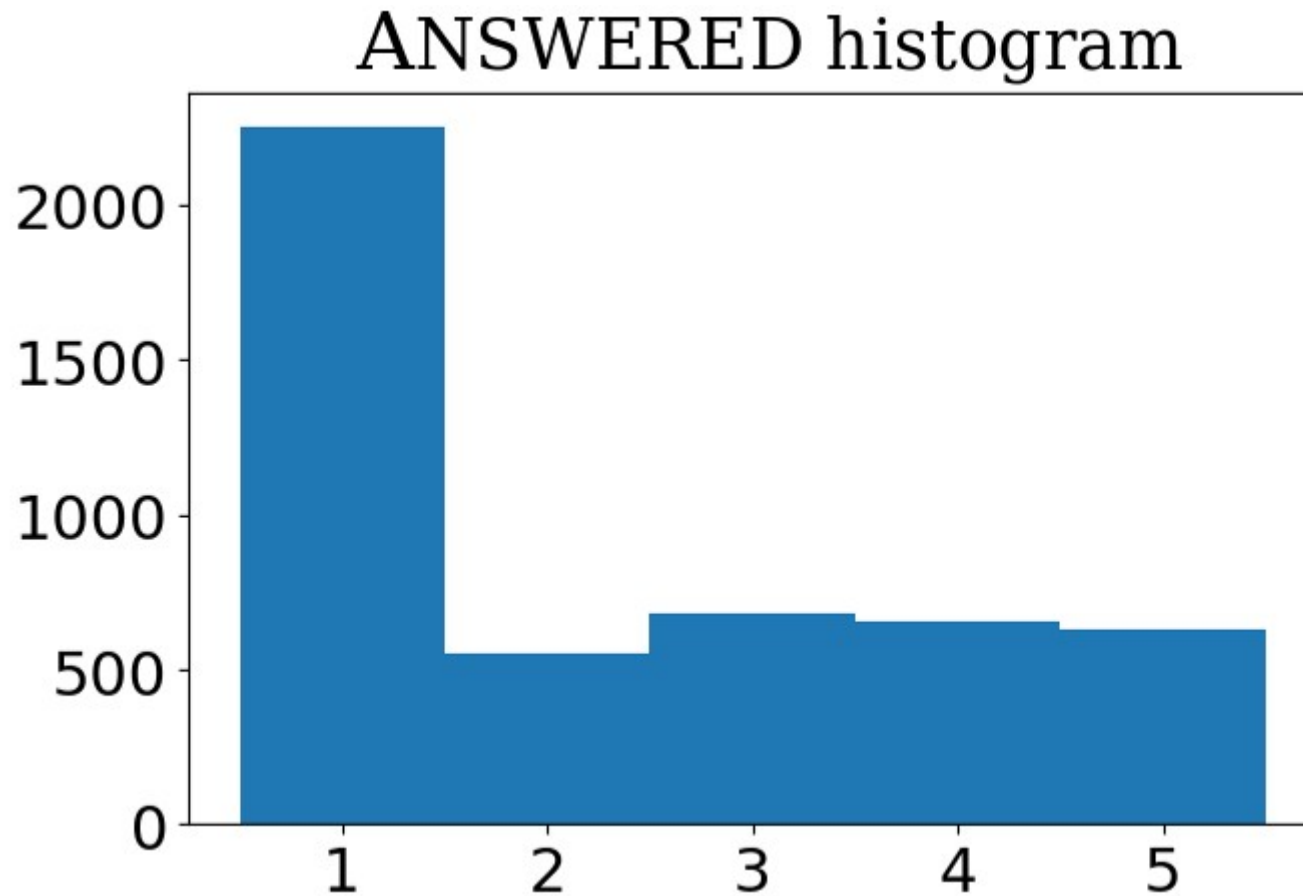
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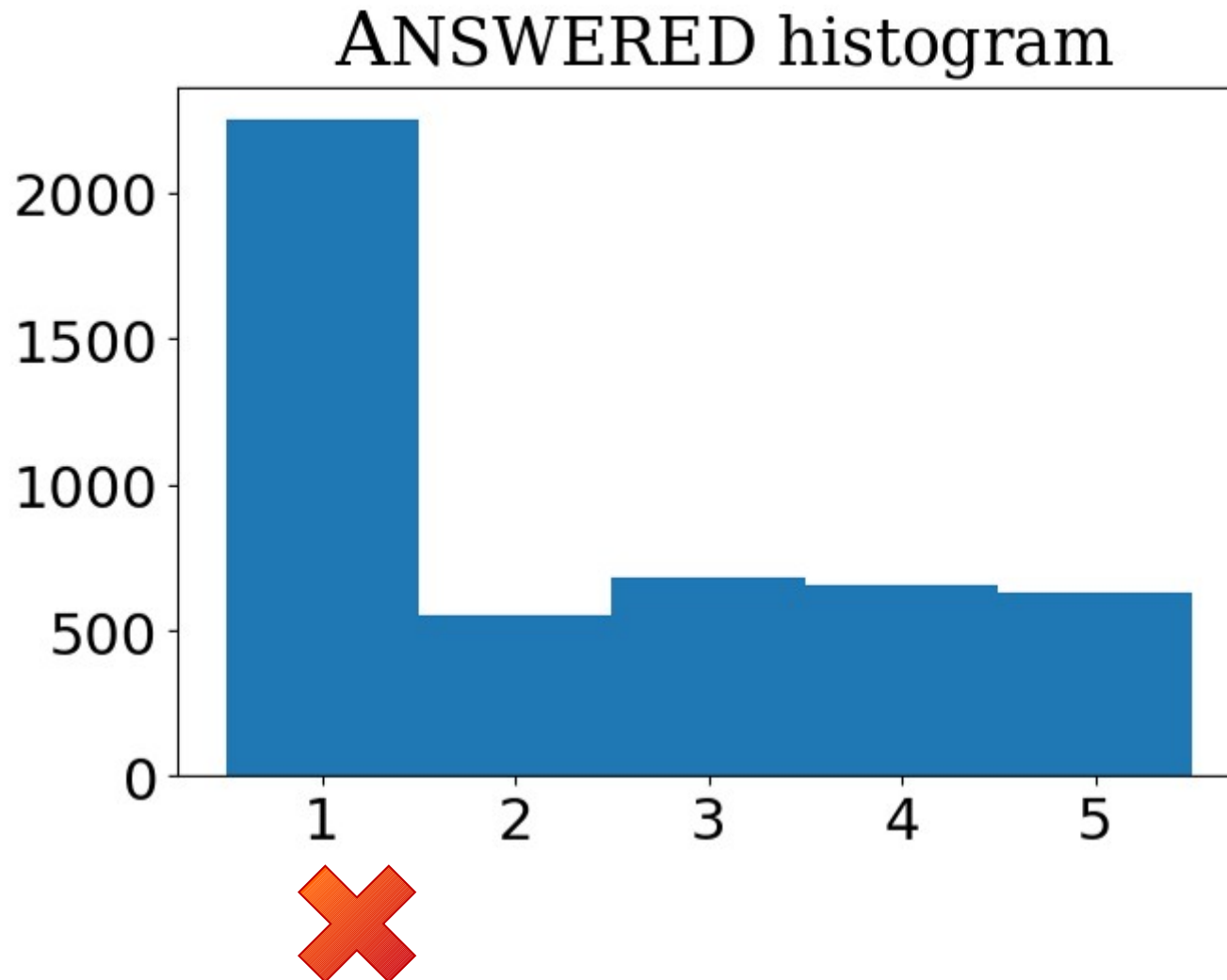
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- **Result:** 863 probes; 4765 questions; 1965 answers.

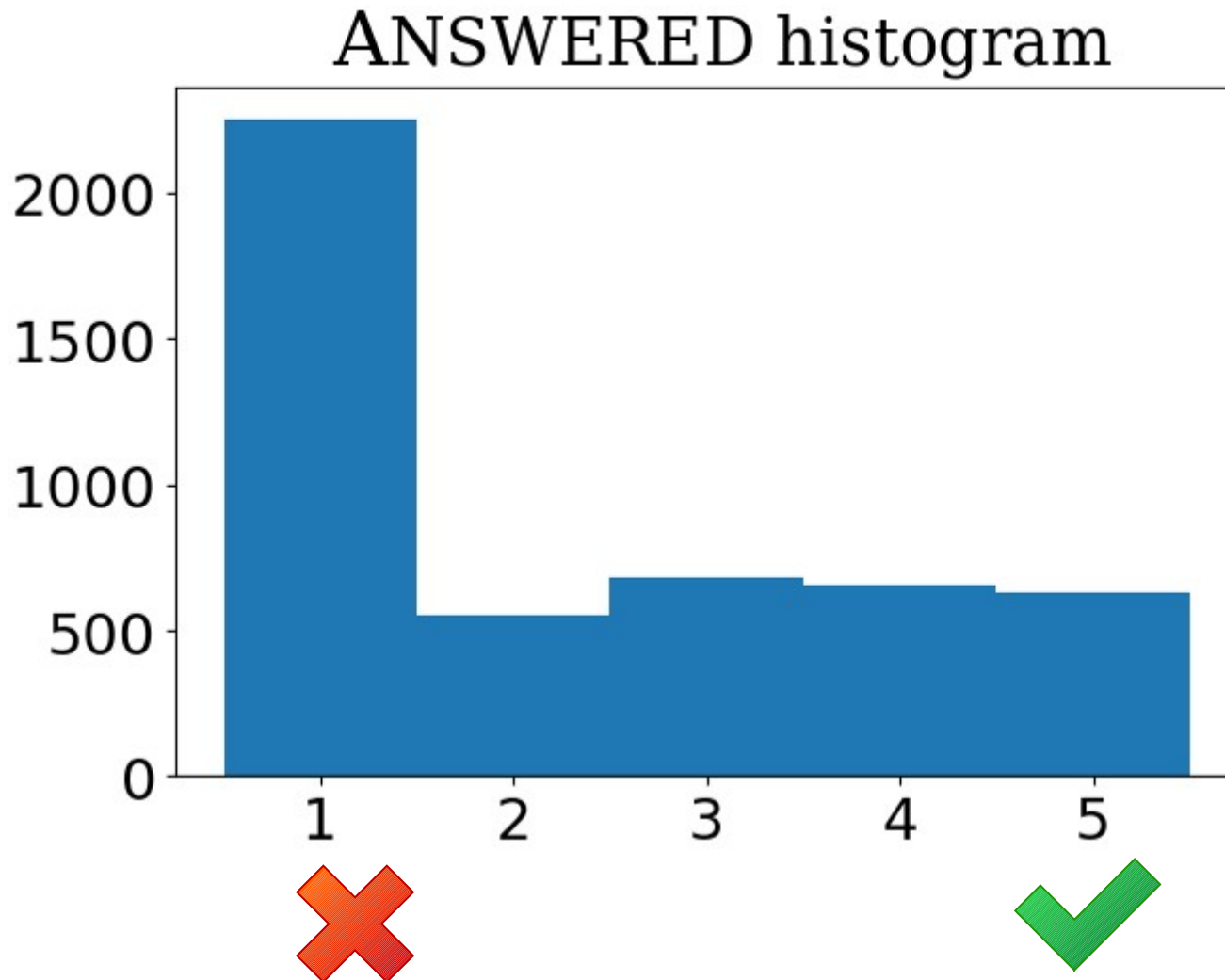
How many questions get answered?



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Mean ANSWERED per genre

Genre	ANSWERED
DISCO-SPICE dialogue	2.11
TED talk presentation	2.50
Constructed story	2.89

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- ANSWERED score of evoked question = ANSWERED score of its best answer (scale 1-5).

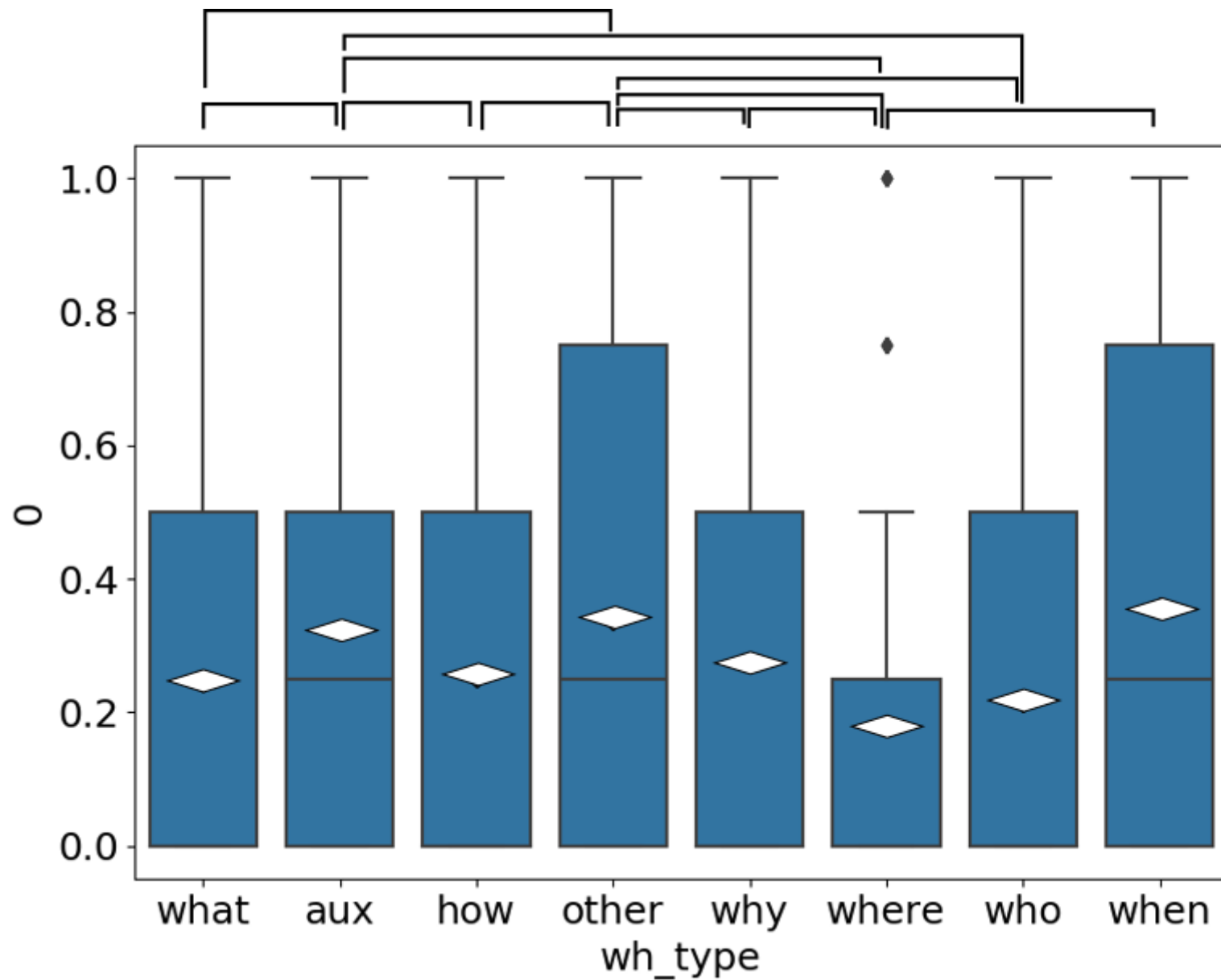
Question types: DISCO-SPICE



Question types: TED-talks



ANSWERED per Q-type: DISCO-SPICE



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No idea?

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Zooming in on the TED-talks

- TED-MDB's existing discourse annotations have better coverage.
DISCO-SPICE covers only within-utterance relations :(
- The TED-talks seem more suitable for our (current) elicitation task anyway.
- So we further annotated and analyzed our TED-portion, and submitted it to LREC as **TED-Q**.

TED-Q + TED-MDB = ♥

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Is discourse structure more explicitly signaled
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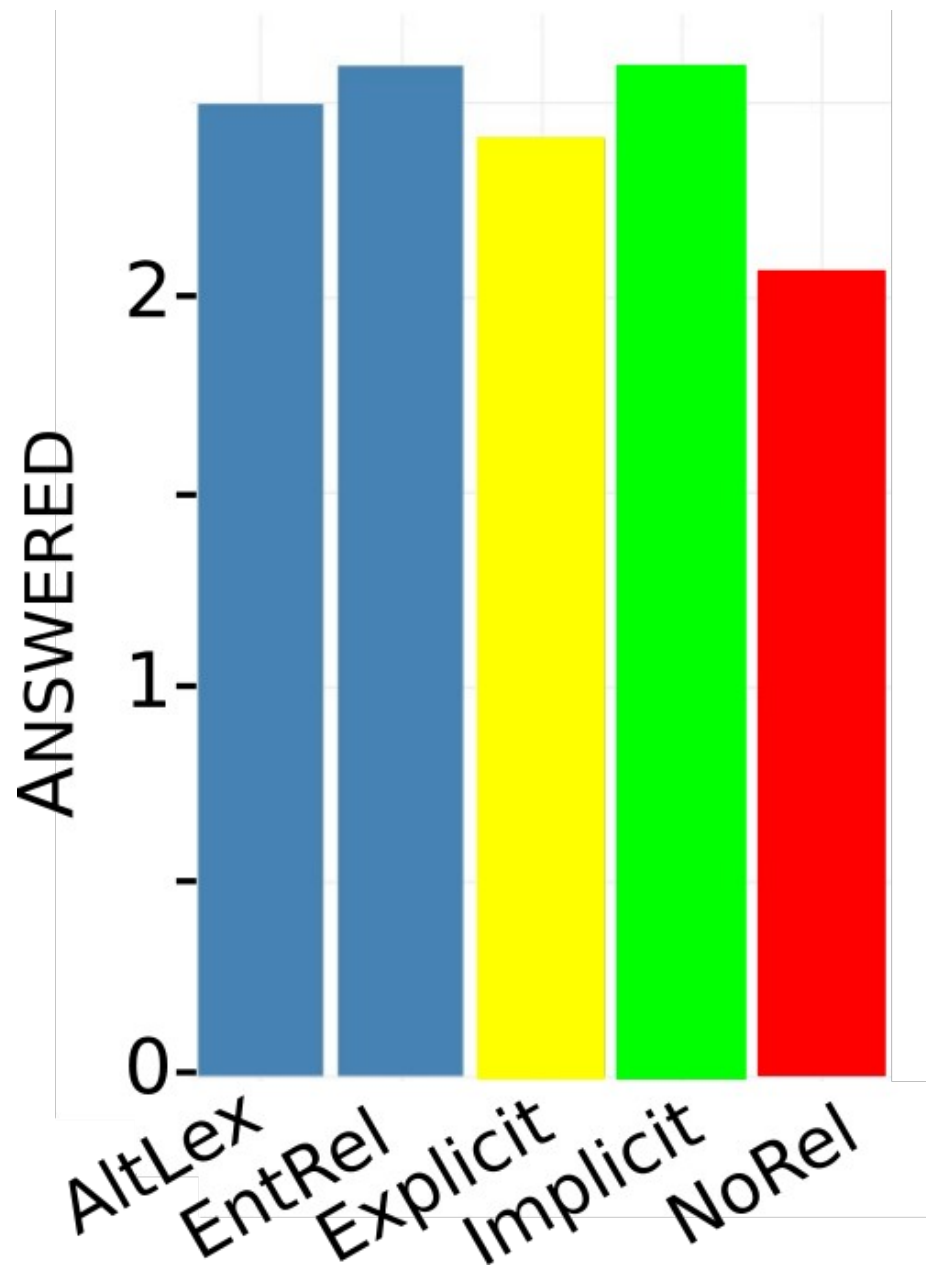
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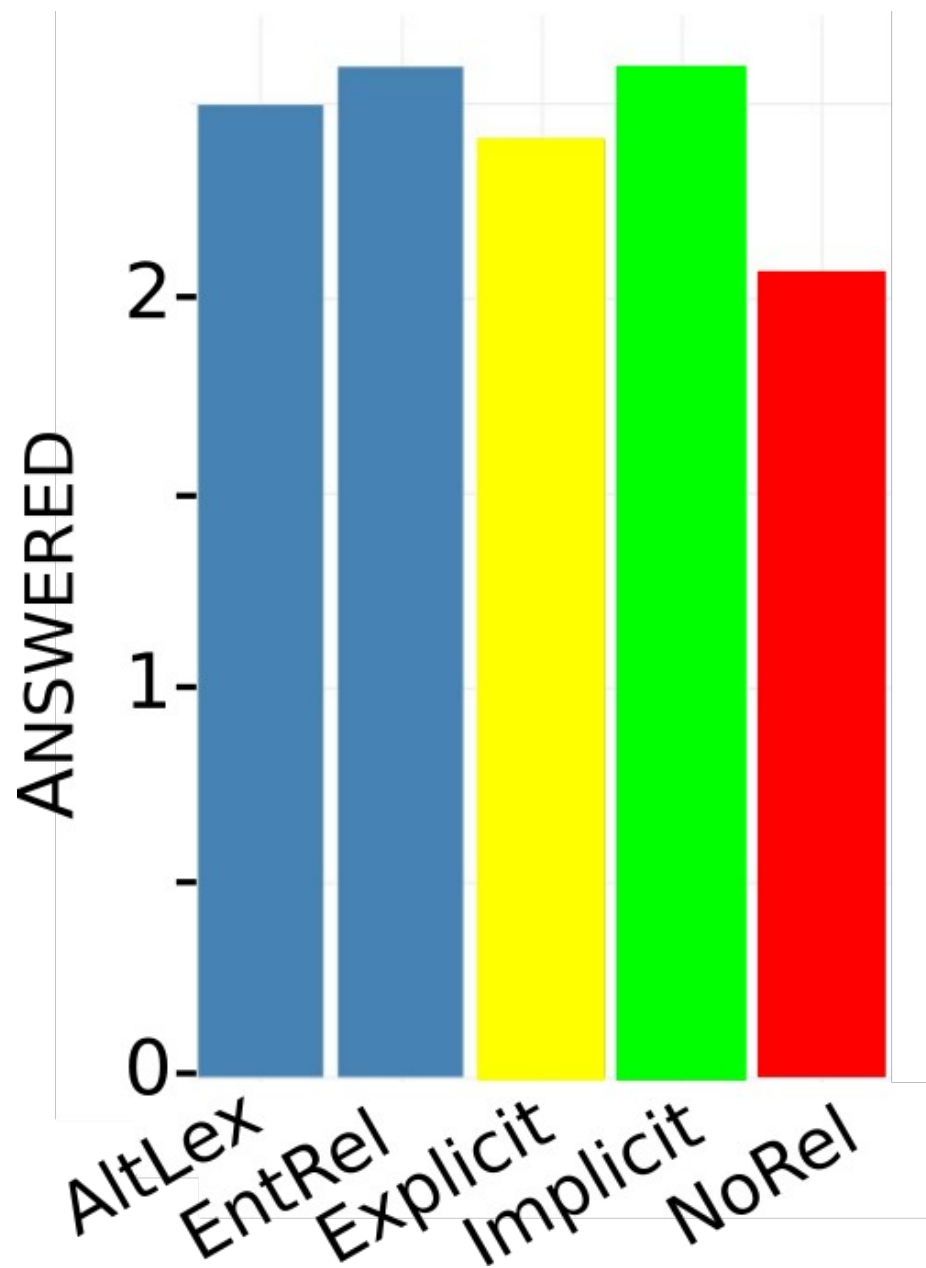
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- ... and whether it is made explicit or not (connectives).

Explicit vs. implicit Discourse Rel.



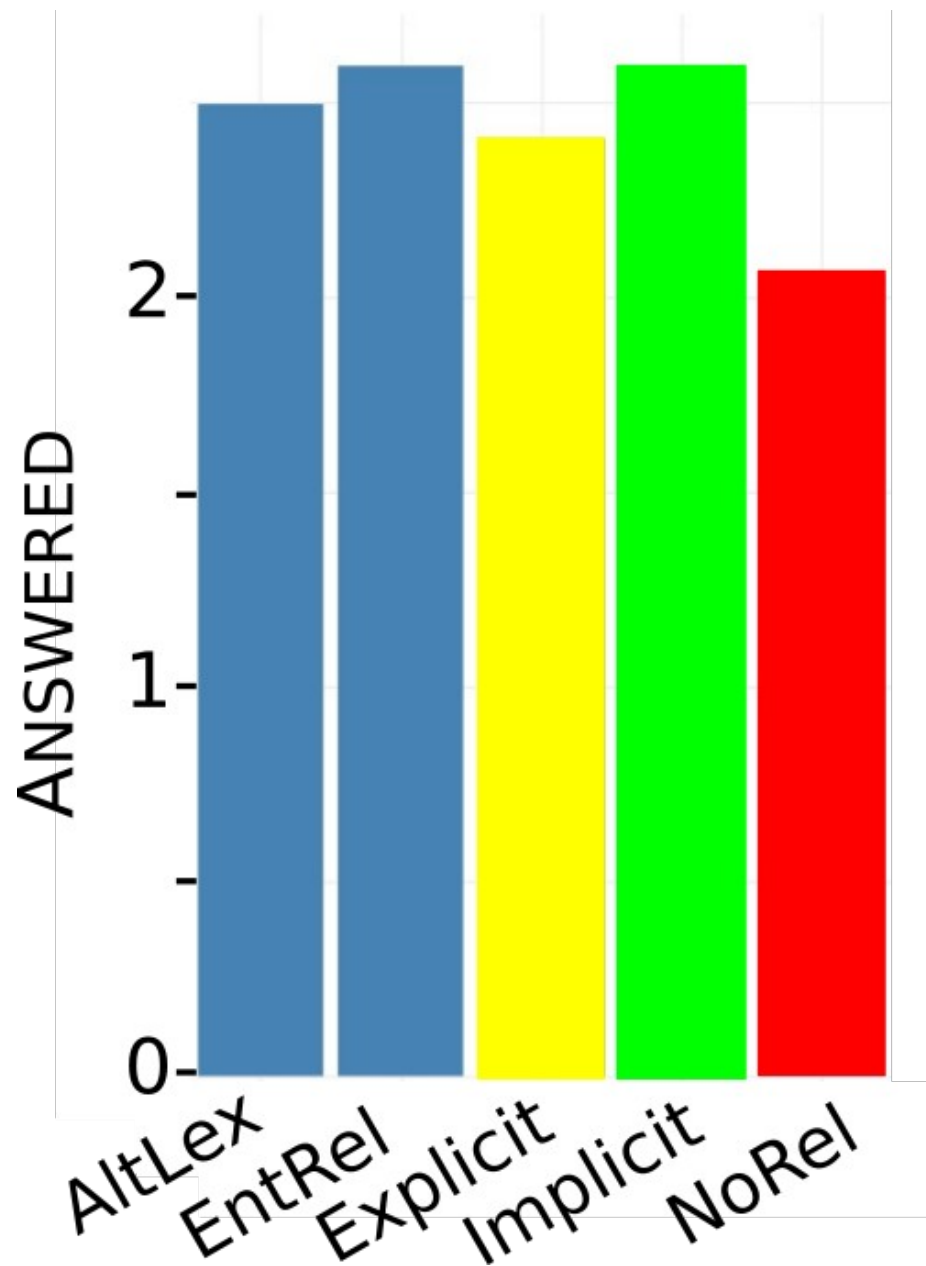
Explicit vs. implicit Discourse Rel.



ANSWERED rating:

- Explicit < Implicit
($t(1580)=2.39, p=.016$)

Explicit vs. implicit Discourse Rel.



ANSWERED rating:

- **Explicit** < **Implicit**
($t(1580)=2.39$, $p=.016$)
- **NoRel** < others
($t(2219)=4.71$, $p<.0001$)

Question types ~ relations?

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No significant effects...



Question types ~ relations?

No significant effects...



← Except here!
12% → 19% (p < .01)

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 - “Causal relations more predictable.”
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- TED-MDB + TED-Q enables a new kind of confirmation:
 - without prior assumptions about predictability;
 - purely data-driven, quantitative, context-sensitive.

Did people pose the same question?

[...] Many of you might be wondering why anyone would choose a life like this, under the thumb of discriminatory laws

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Why would people want to live this way?

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Why would anyone live such a dangerous life?

Why are these laws like this?

Why choose homelessness?

Annotating question relatedness

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- Crowdsourced task:

<https://workersandbox.mturk.com/projects/3L2A3M5C0728O1QYNC4O7LNJPOCGQW/tasks>

Annotating question relatedness

► Please read the snippet:

[...] We can still see the little moon, but let's go ten times further away again. Here we are at the edge of the solar system, out at the Kuiper Belt.

► Next, compare the questions it evoked:

	Questions:	How related are target (T) and comparison (C) question?				
Target (T):	Is the Kuiper belt similar to the asteroid belt?					
Comparison (C):	What is the Kuiper Belt?					?
Comparison (C):	What is the Kuiper Belt?					?
Comparison (C):	Can you see the edge of the solar system with a telescope?					?
Comparison (C):	What do we see from the Kuiper Belt?					?

Our resulting dataset: TED-Q

Elicitation phase:	Comparison phase:
texts: 6	question pairs: 4516
words: 6975	participants/pair: 6
probe points: 460	participants: 163
participants/probe: 5+	judgments: 30412
participants: 111	RELATED mean: 1.21
questions: 2412	RELATED std: 0.79
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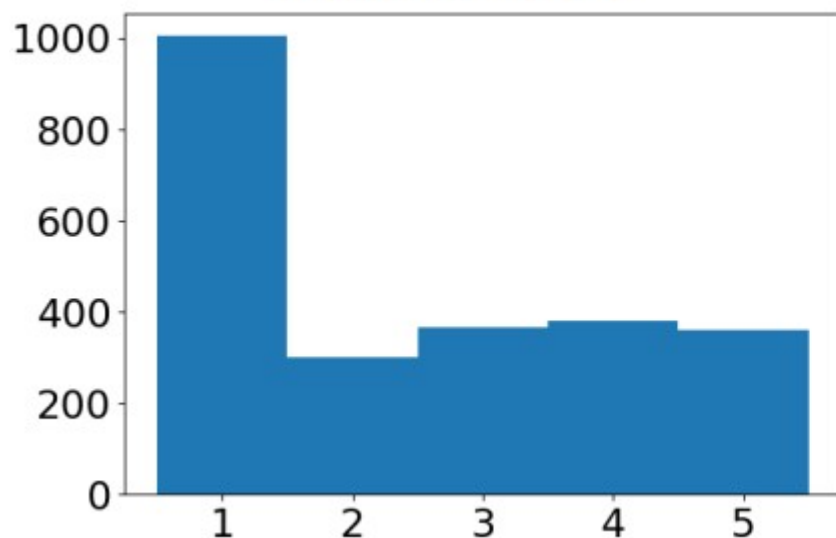
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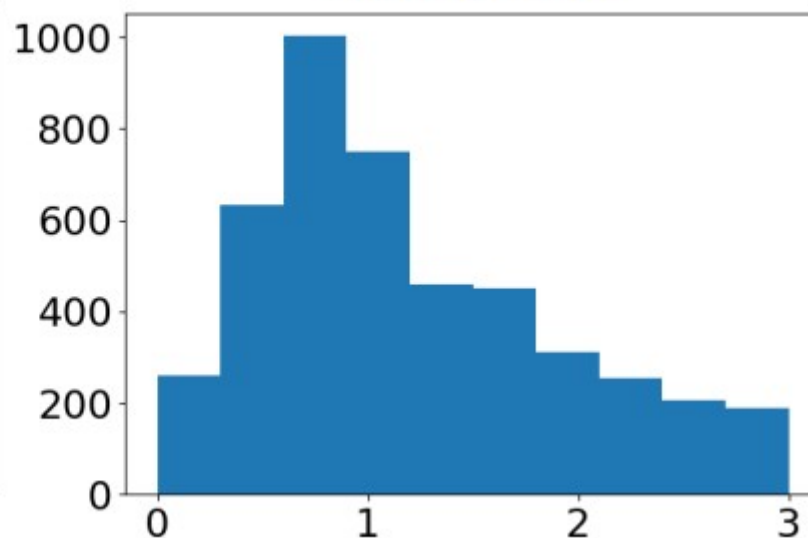
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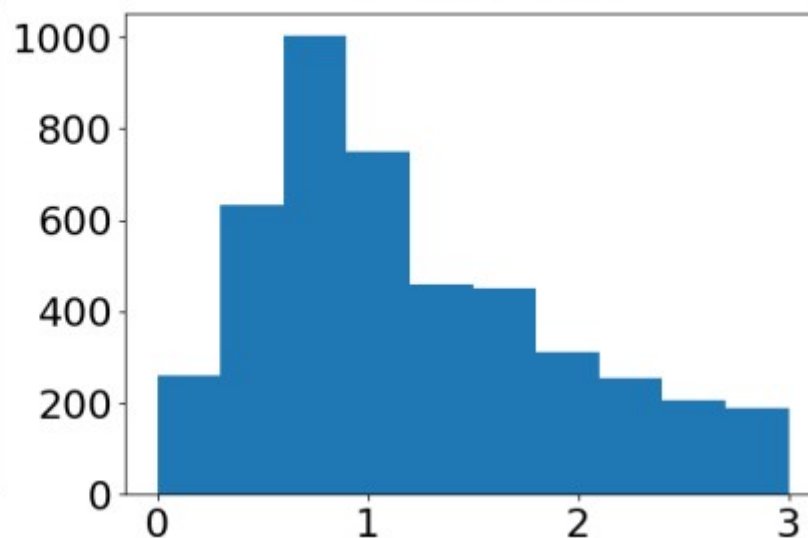
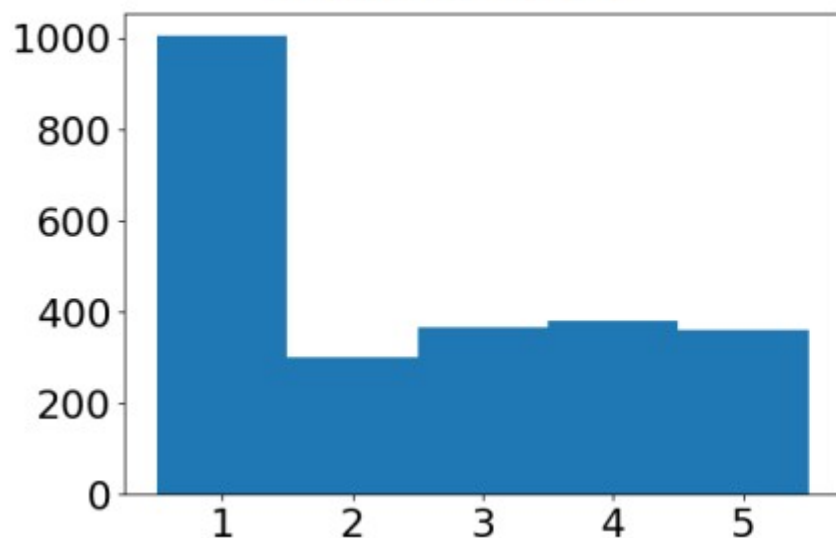
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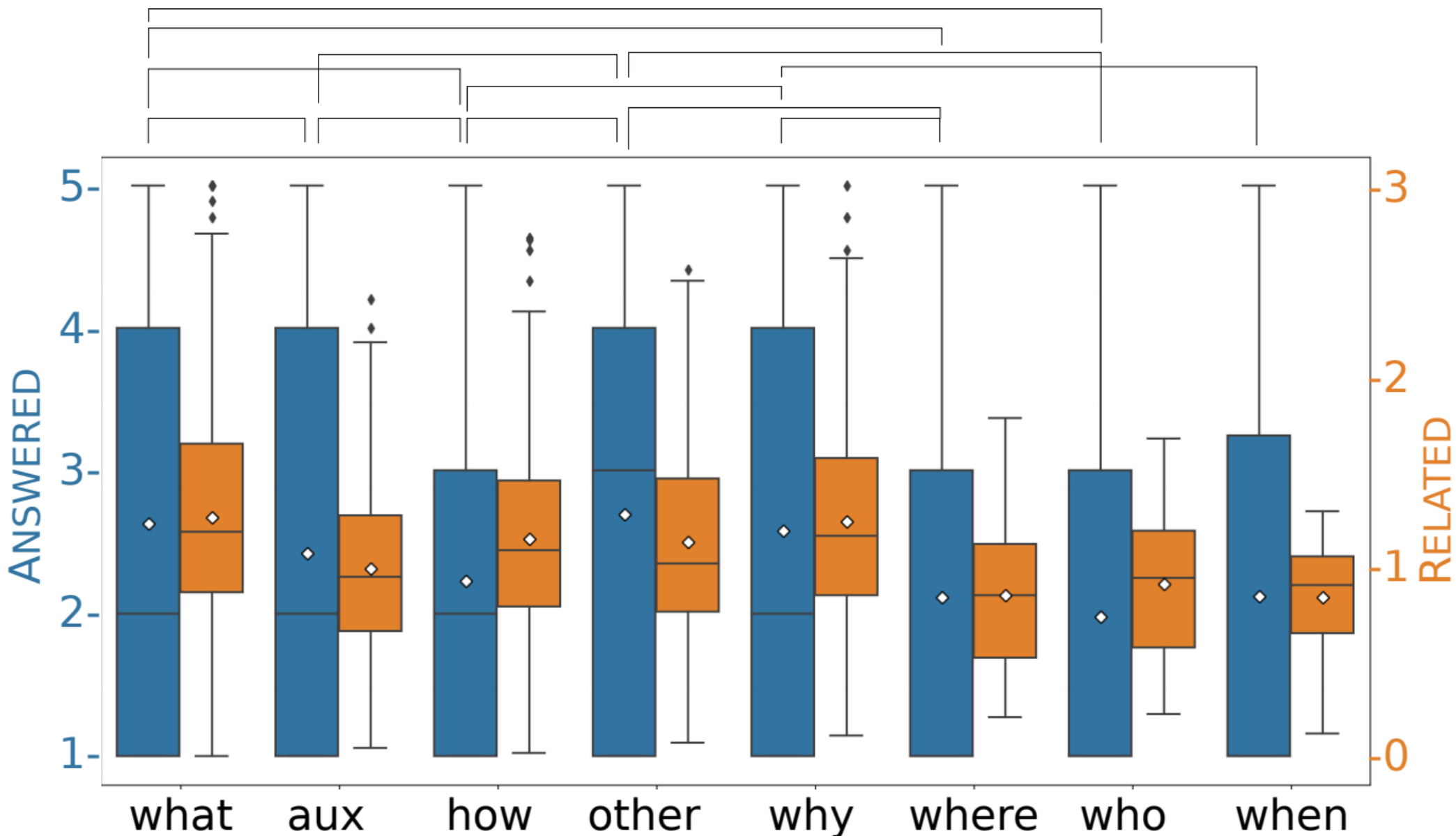
ANSWERED ← Spearman 0.17 → **RELATED**



ANSWERED & RELATED per wh-type



ANSWERED & RELATED per wh-type



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