

Foundations: How to define and measure “understanding”

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Seminar “What do language models really understand”?

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Plan for today

- The octopus thought experiment and referentialism
- Responses:
 - Opportunities for grounding
 - Alternative views of what it means to understand
- Methods for evaluating understanding abilities
 - Benchmarks
 - Behavioral experiments
 - Probing
- Guidelines for presentations, reading papers, and commentaries

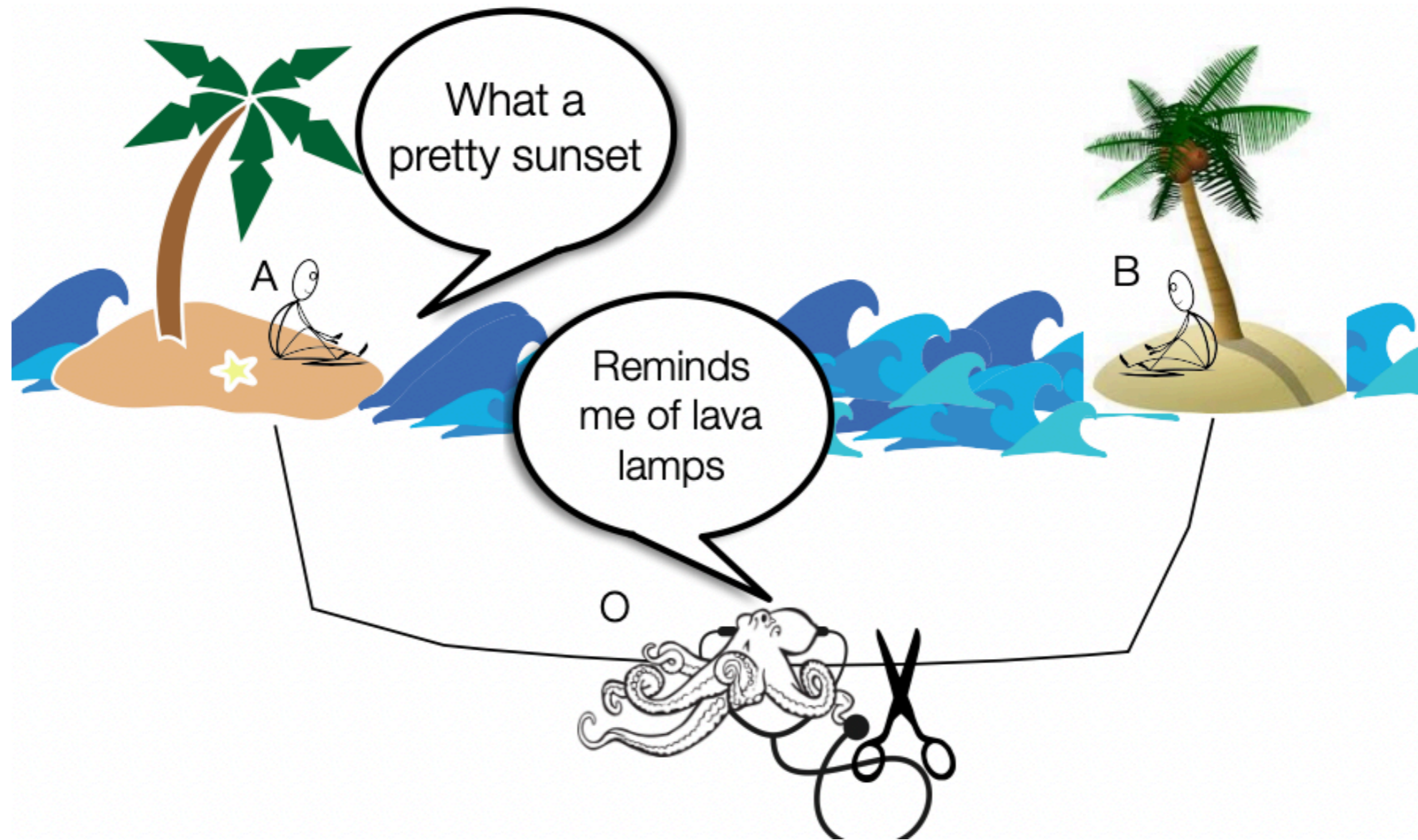
What is understanding?

- One view:
 - We don't just use language for fun — we use language to achieve **communicative intents**
 - Formalization:

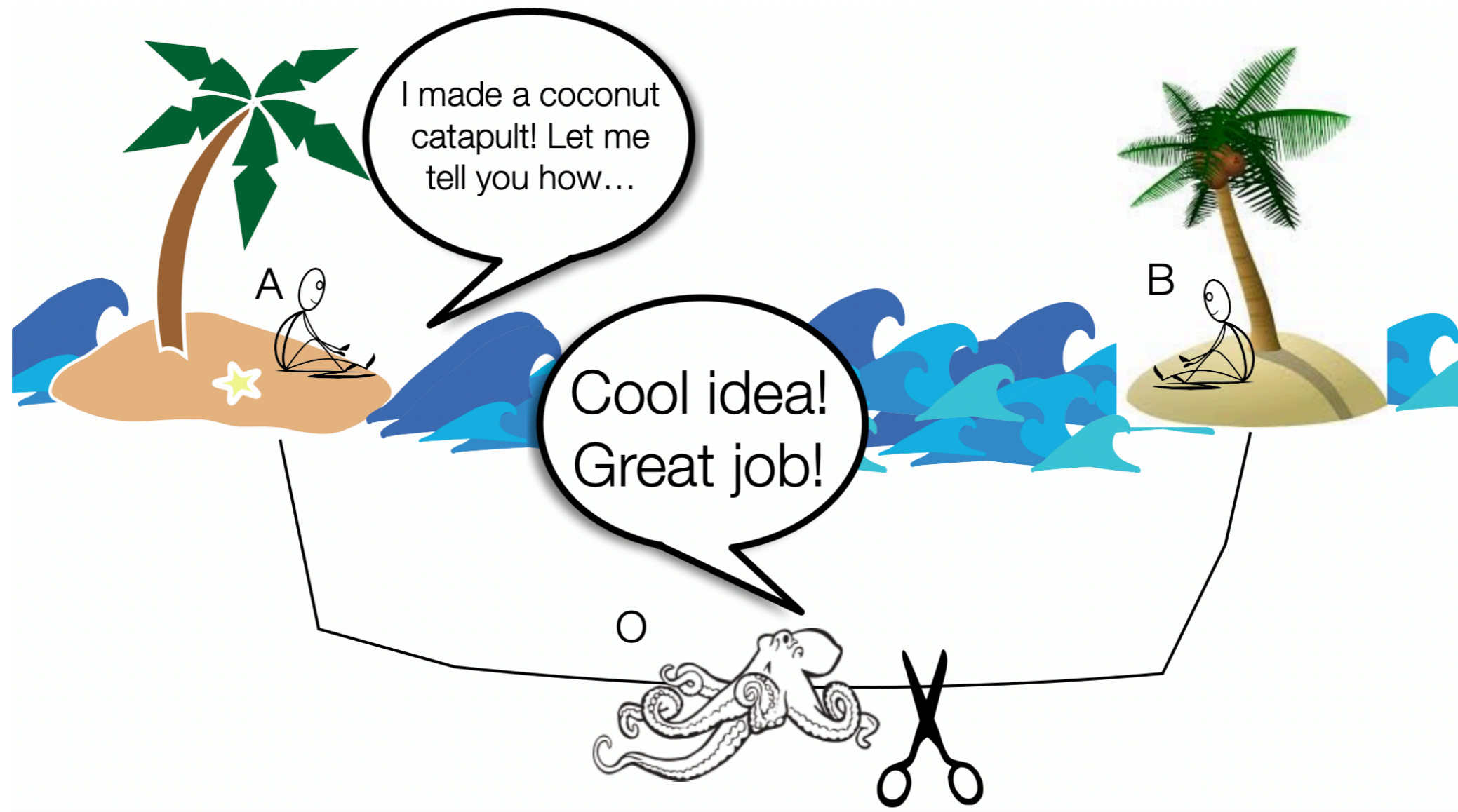
Meaning: $M \subseteq E \times I$ (relation between natural language expressions e and communicative intents i)

- Communicative intents are **something outside of language and grounded in the real world**
- **Understanding:** given an expression e , in a context, recover the communicative intent i

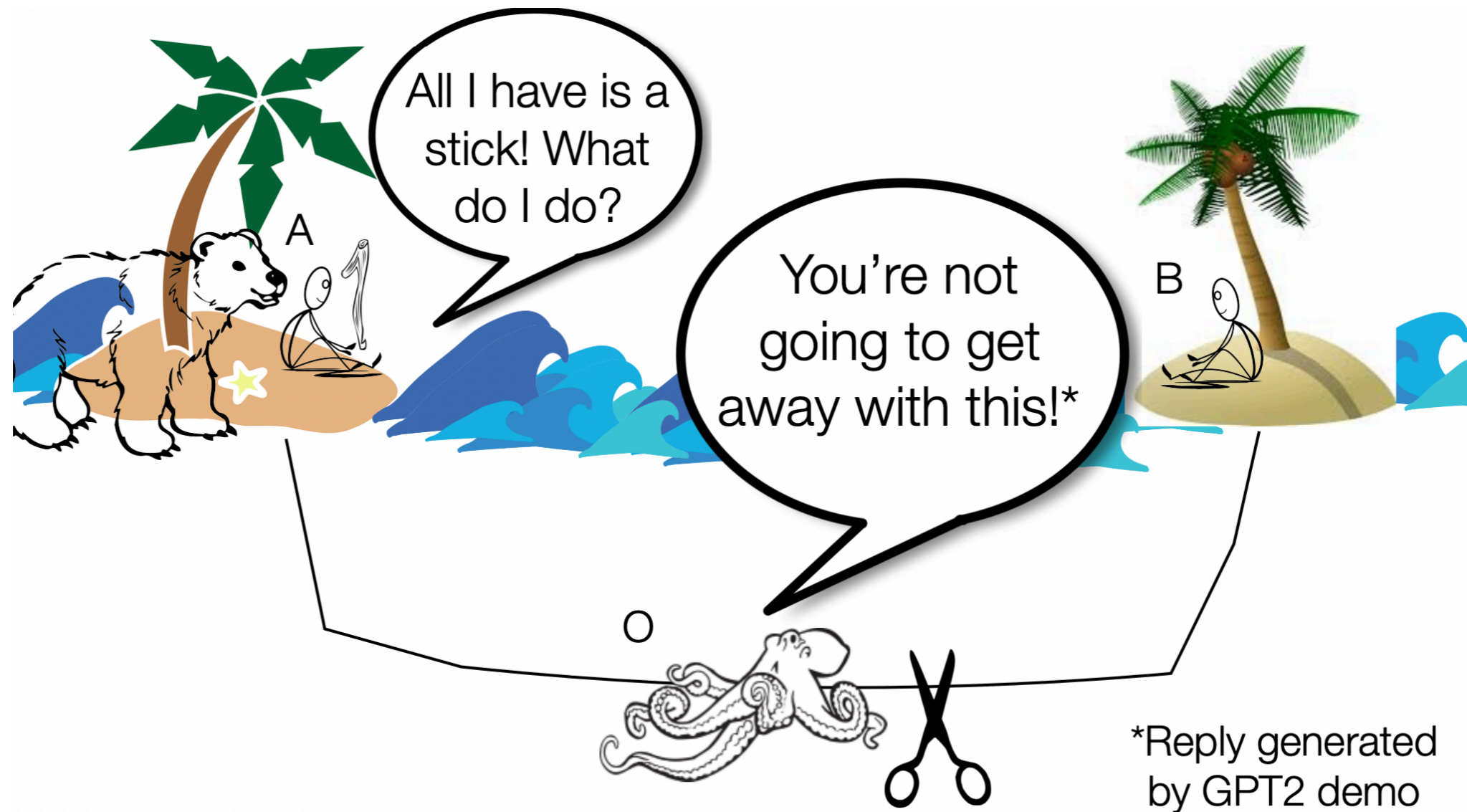
The octopus test



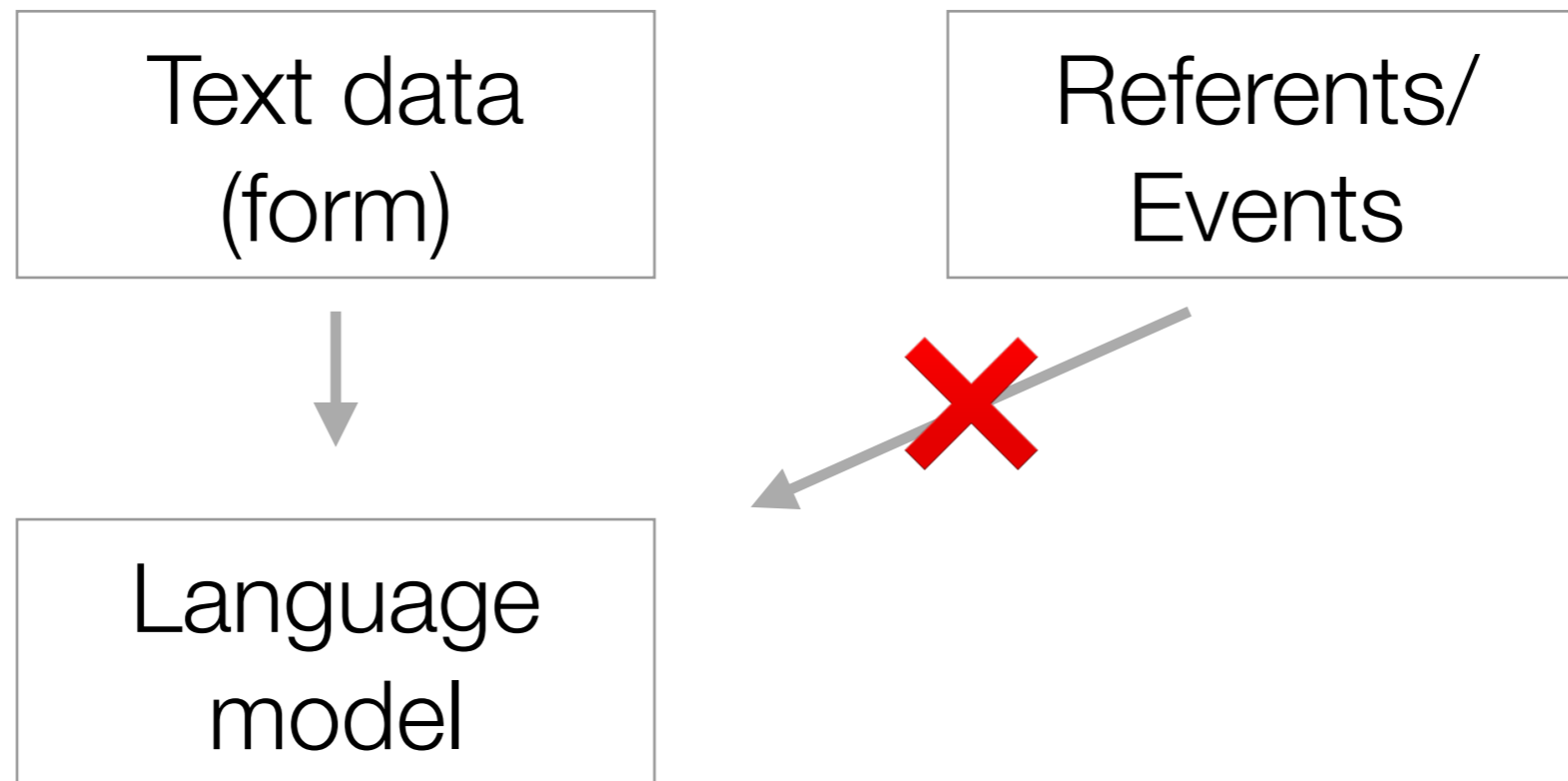
The octopus test



The octopus test



Can language models understand?



Responses

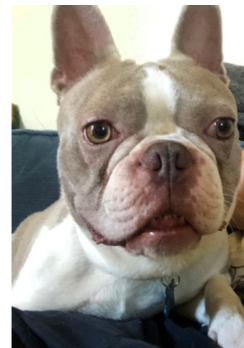
- Some grounding may happen when training only on form
 - e.g., unit tests in code
- Still relevant? Best LLMs are grounded in several ways (how?)
- Under specific assumptions about language use, pure LMs can learn whether one statement entails another statement (Merrill et al., 2022)
- There are alternative views of “understanding” than the one expressed by Bender & Koller.

Bender and Koller's view: Referentialism

- **Referentialism:**

- Words and phrases **map to entities and events** in the real world
- An agent understands language if it is able to do this **mapping** and to **evaluate whether statements are true in the world**

Archie ↔



Archie is a dog ↔ True

One alternative view: Pragmatism

- **Pragmatism**

- What matters is that the agent be disposed **to use language in the right way**
- This may include appropriate inference and reasoning patterns, appropriate conversational moves, etc.
- **Being able to use language in the right way** constitutes understanding

Archie is a dog \longrightarrow Archie is a mammal

Agent1: What day of the week is it today?

Agent2: It is Thursday.

Can language models understand?

- Under a pragmatist view: Maybe?



What day of the week is it today?



Today is Wednesday.

Methods for assessing understanding abilities

Methods for assessing understanding abilities

- Task benchmarks (e.g., Natural Language Inference benchmarks)
- Behavioral experiments (aka “Targeted evaluations”)
- Probing

Benchmarks

- A classic benchmark:
 - Crowdsourced examples
 - Randomly split into training/development/test examples
 - Model is trained on training split and evaluated on test split resulting in an overall accuracy score

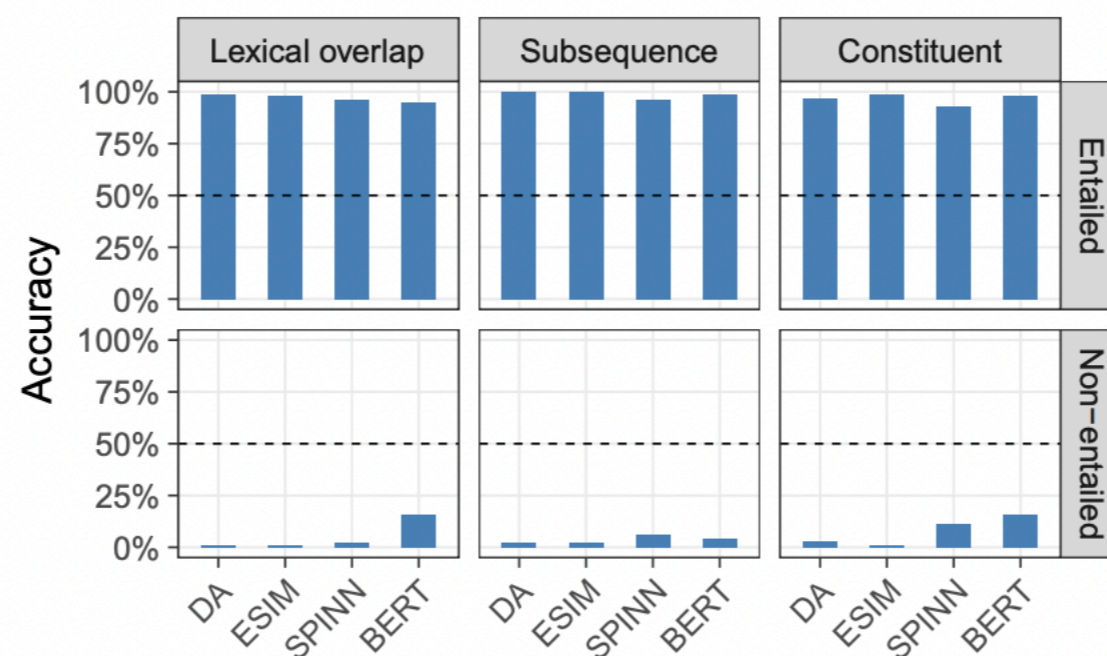
Issues with benchmarking

- Statistical artifacts in SNLI:

Premise	A woman selling bamboo sticks talking to two men on a loading dock.
Entailment	There are at least three people on a loading dock.
Neutral	A woman is selling bamboo sticks to help provide for her family.
Contradiction	A woman is not taking money for any of her sticks.

Issues with benchmarking

Heuristic	Definition	Example
Lexical overlap	Assume that a premise entails all hypotheses constructed from words in the premise	The doctor was paid by the actor. ————→ The doctor paid the actor. WRONG
Subsequence	Assume that a premise entails all of its contiguous subsequences.	The doctor near the actor danced. ————→ The actor danced. WRONG
Constituent	Assume that a premise entails all complete subtrees in its parse tree.	If the artist slept , the actor ran. ————→ The artist slept. WRONG



(b)

Issues with benchmarking

- Tasks are usually quite general
 - Question answering
 - Natural language inference
 - ...
- **Difficult to identify systematic shortcomings**

Behavioral experiments / targeted evaluation suites

- Inspired by psycholinguistics experiments
- Small test sets that target a specific behavior, e.g., negation
- Models are usually not trained on similar examples
 - Evaluates out-of-distribution examples

Example: Evaluating whether models learned dependencies necessary for reflexives

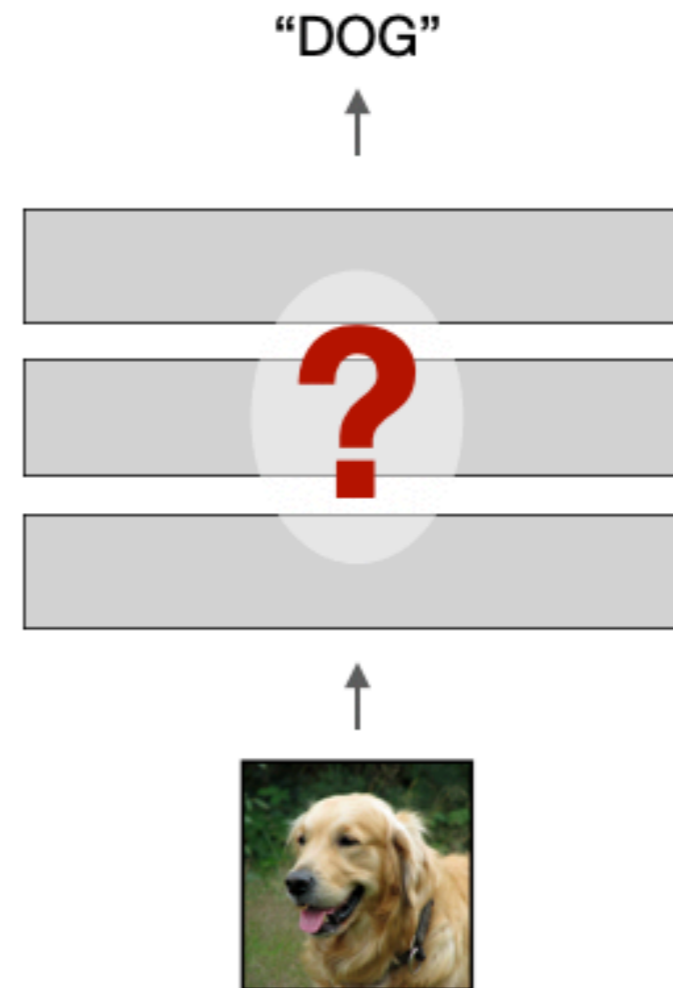
The bankers who the pilot embarrassed hurt _____

$P(\text{themselves} \mid \textit{Context}) > P(\text{herself} \mid \textit{Context})?$

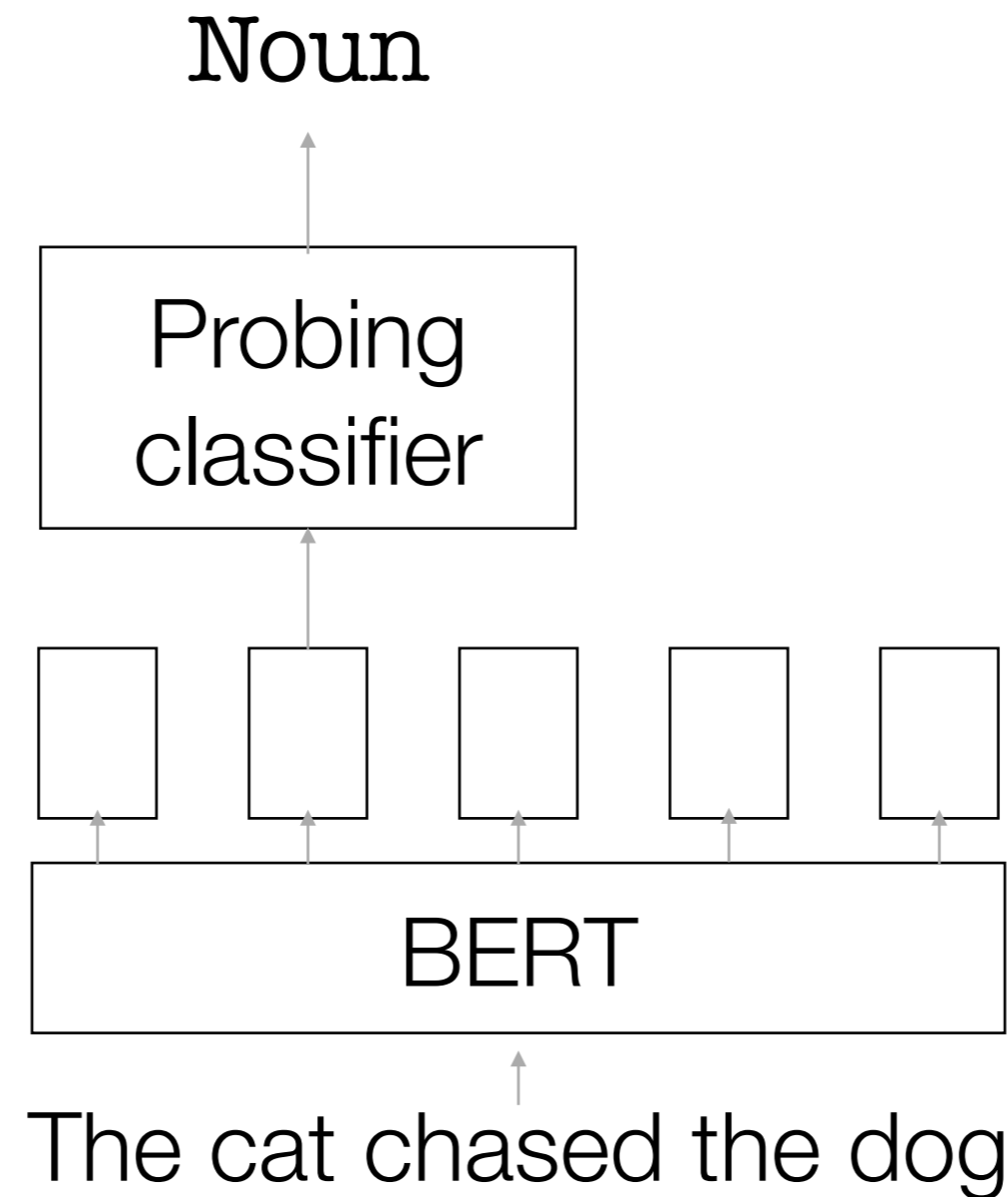
The bankers thought the pilot embarrassed _____

$P(\text{herself} \mid \textit{Context}) > P(\text{themselves} \mid \textit{Context})?$

Probing



Example: determining whether representations encode something about part-of-speech tags



Takeaways

- Two prominent views of what it means to understand:
 - Referentialism and Pragmatism
- Methods for evaluating abilities of language models
 - Benchmarks
 - Targeted evaluations
 - Probing

Guidelines for readings and presentations

Questions to keep in mind while doing the readings

- What are the properties of the model(s) being used?
 - Autoregressive model? Masked LM? Something else?
 - How was the model trained? Additional training objectives on top of LM task?
- How was the understanding ability evaluated? Did the evaluation task potentially provide additional supervision?
- What kind of data was being used? Naturalistic? Hand-crafted? Can we rule out statistical artifacts in the data? Could the model have learned shortcuts?
- Do the authors talk about “understanding”? If so, what kind of definition of “understanding” do they (seem to) assume?
- Does the paper report results from models of different size? Does size seem to matter for the evaluated ability?

Guidelines for presentations

- Length: 15-25min + 15-25min of discussion
- Contents:
 - Summary of the **main questions, methods and results**
 - (optional) Background on model and data
 - Discussion of the strengths and weaknesses of the paper
- Slides and handouts are optional

Guidelines for weekly comments

- Superficial questions/comments:
 - Questions/comments that could have been written by reading just the abstract or a paragraph of the paper
 - “I didn’t understand X...” (fine to mention that as well but not as the only comment!)
- Examples of insightful questions/comments:
 - Connect multiple points made in the paper
 - Relate findings of a paper to other papers we’ve read
 - Relate to some of the higher-level questions we are asking in this course