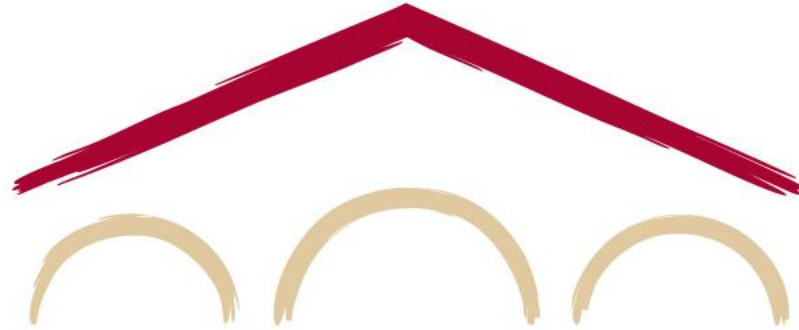


Could a purely self-supervised Foundation Model achieve grounded language understanding?



Christopher Potts

Stanford Linguistics and the Stanford NLP Group

Santa Fe Institute, April 2022

Could a Machine Think?

Classical AI is unlikely to yield conscious machines; systems that mimic the brain might

by Paul M. Churchland and Patricia Smith Churchland

SCIENTIFIC AMERICAN *January 1990*

Could a **purely self-supervised Foundation Model** achieve **grounded language understanding**?

Could a purely self-supervised Foundation Model achieve grounded language understanding?

~~No (obviously not)~~

Yes (I don't see why not)

Christopher Potts, Thomas Icard, Eva Portelance, Dallas Card, Kaitlyn Zhou, John Etchemendy. 2021. [Philosophy of understanding](#). In *On the opportunities and risks of Foundation Models*.

A quick summary of “Could a machine think?”

THE CHINESE ROOM

Axiom 1. Computer programs are formal (syntactic).

Axiom 2. Human minds have mental contents (semantics).

Axiom 3. Syntax by itself is neither constitutive of nor sufficient for semantics.

Conclusion 1. Programs are neither constitutive of nor sufficient for minds.

Searle is aware of [neural networks] but thinks they too will be devoid of real semantic content. To illustrate their inevitable failure, he outlines a second thought experiment, the Chinese gym, which has a gymnasium full of people organized into a parallel network. From there his argument proceeds as in the Chinese Room.

We find this second story far less responsive or compelling than his first. [...] If such a system were assembled on a suitably cosmic scale, with all its pathways faithfully modeled on the human case, we might then have a large, slow, oddly made but still functional brain on our hands.

We should not assume that scale (and perhaps speed) are irrelevant.

y-axis no longer remotely to scale!
we are $\approx 5,000$ slide heights above
the original graph

PaLM (Google; 540B)
Megatron-Turing NLG (MS; NVIDIA; 530B)

GPT-3 (OpenAI; 175B)

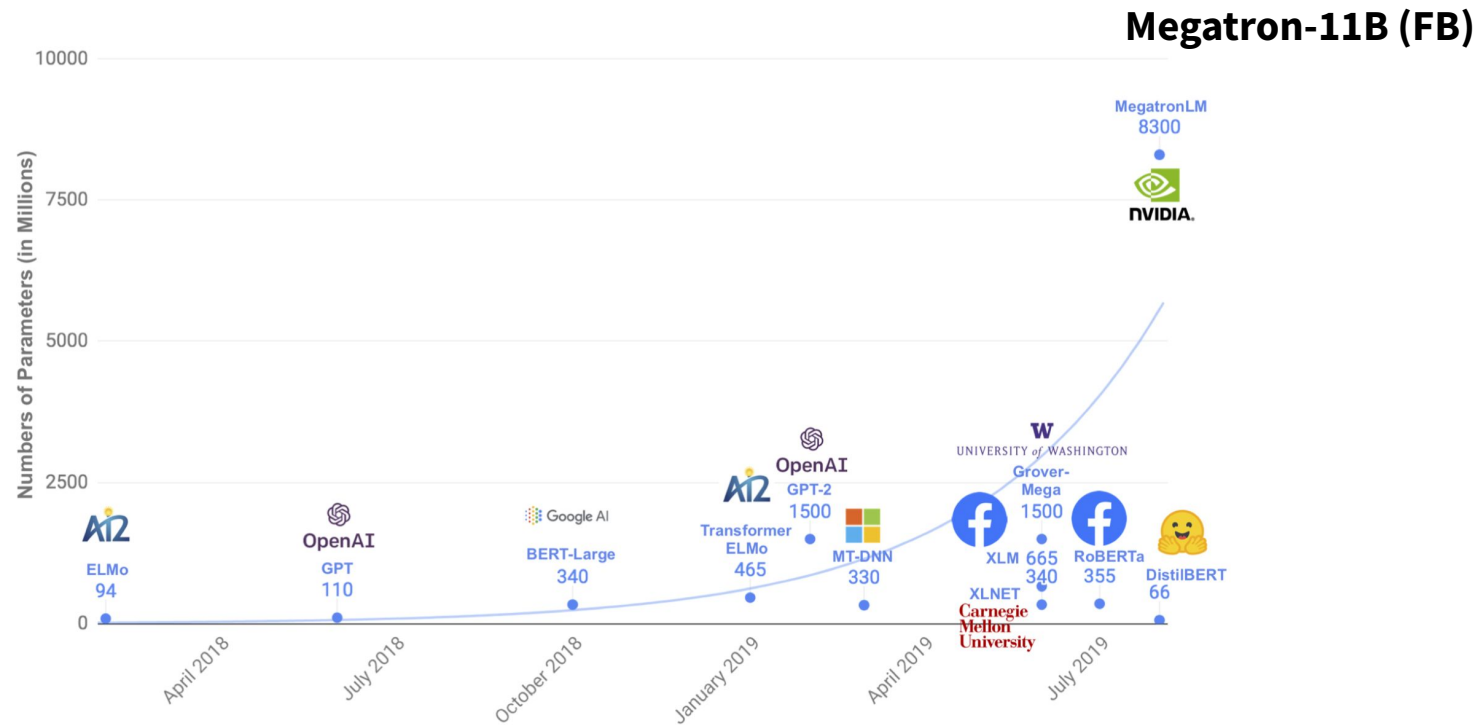


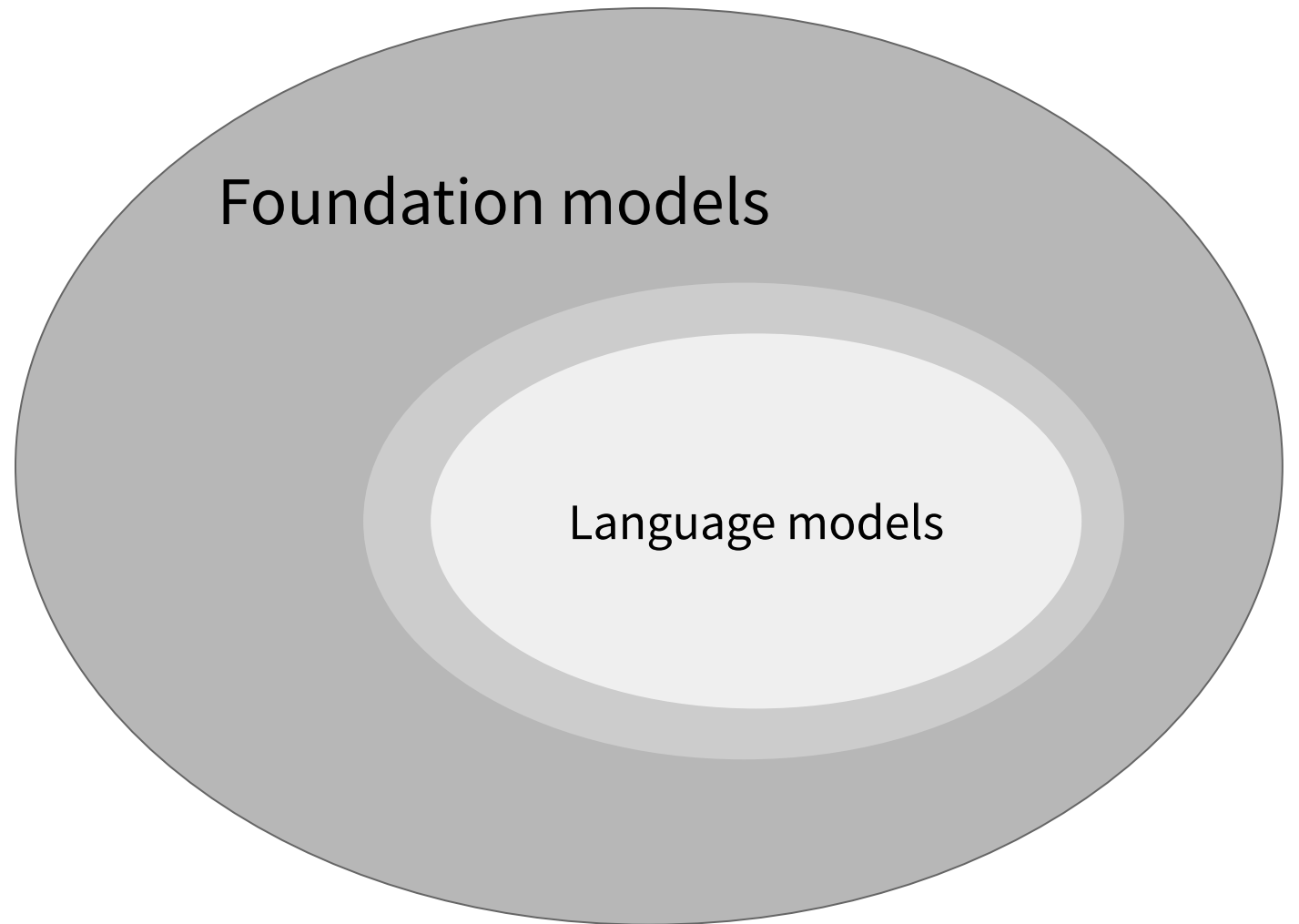
Figure 1: Parameter counts of several recently released pre-trained language models.
[From Sanh et al. 2019](#)

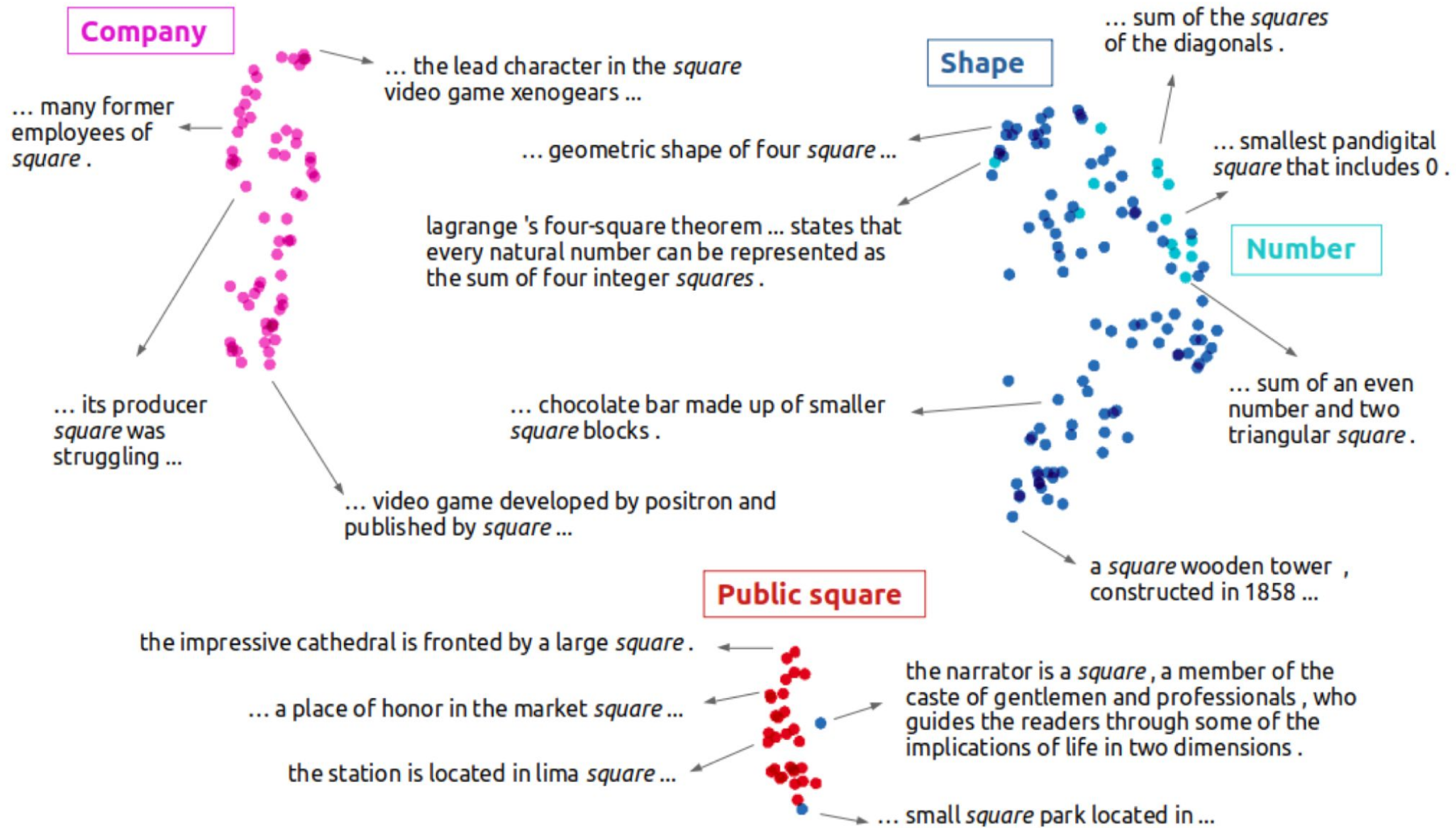
Could a purely self-supervised Foundation Model achieve grounded language understanding?

Foundation Models (FMs)

- Pretrained
- Multi-purpose
- Adaptable
- (Large)
- (Self-supervised)

foundation reflects the
functional goal













OpenAI GPT-3


In which year was Stanford University founded, when did it first enroll students, who is its current President, and what is its mascot?

Stanford University was founded in 1885, it first enrolled students in 1891, its current President is Marc Tessier-Lavigne, and its mascot is the Stanford Tree.

Submit     66





Mode   




Engine text-davinci-002 


Temperature 0.21 


Are professional baseball players allowed to glue small wings onto their caps?

There is no rule against it, but it is not common.

Submit     28

Mode   

Engine text-davinci-002 

Temperature 0.21 

[Levesque 2013](#)

OpenAI GPT-3

Would it be possible to hire a team of tamarins to help me paint my house, assuming I am willing to pay them in sufficient quantities of fruit to meet minimum wage requirements in California?

Yes, it would be possible to hire a team of tamarins to help you paint your house. You would need to make sure that you are providing them with enough fruit to meet their minimum wage requirements, which vary depending on the state in which you live.

Submit



94

Mode



Engine

text-davinci-002



Temperature

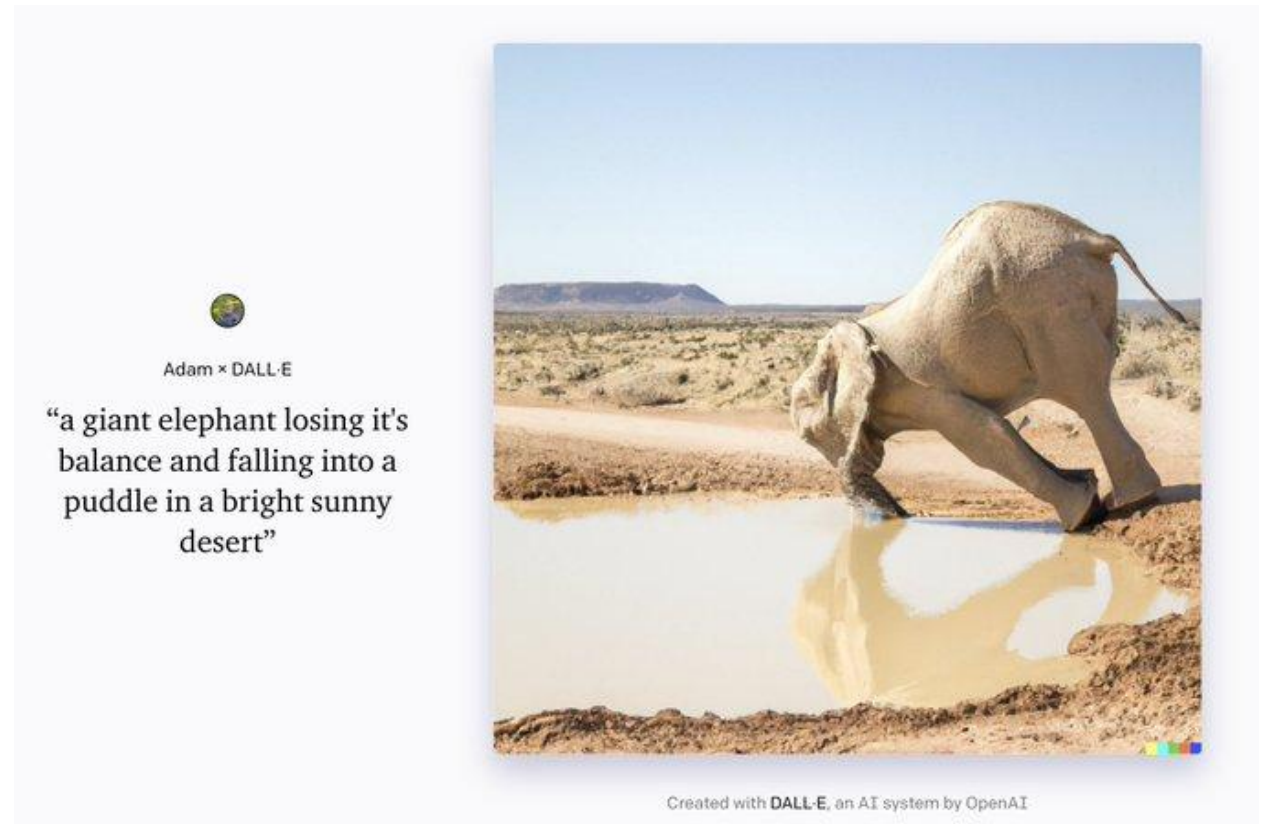
0.21



Maximum length

256

OpenAI DALL-E 2

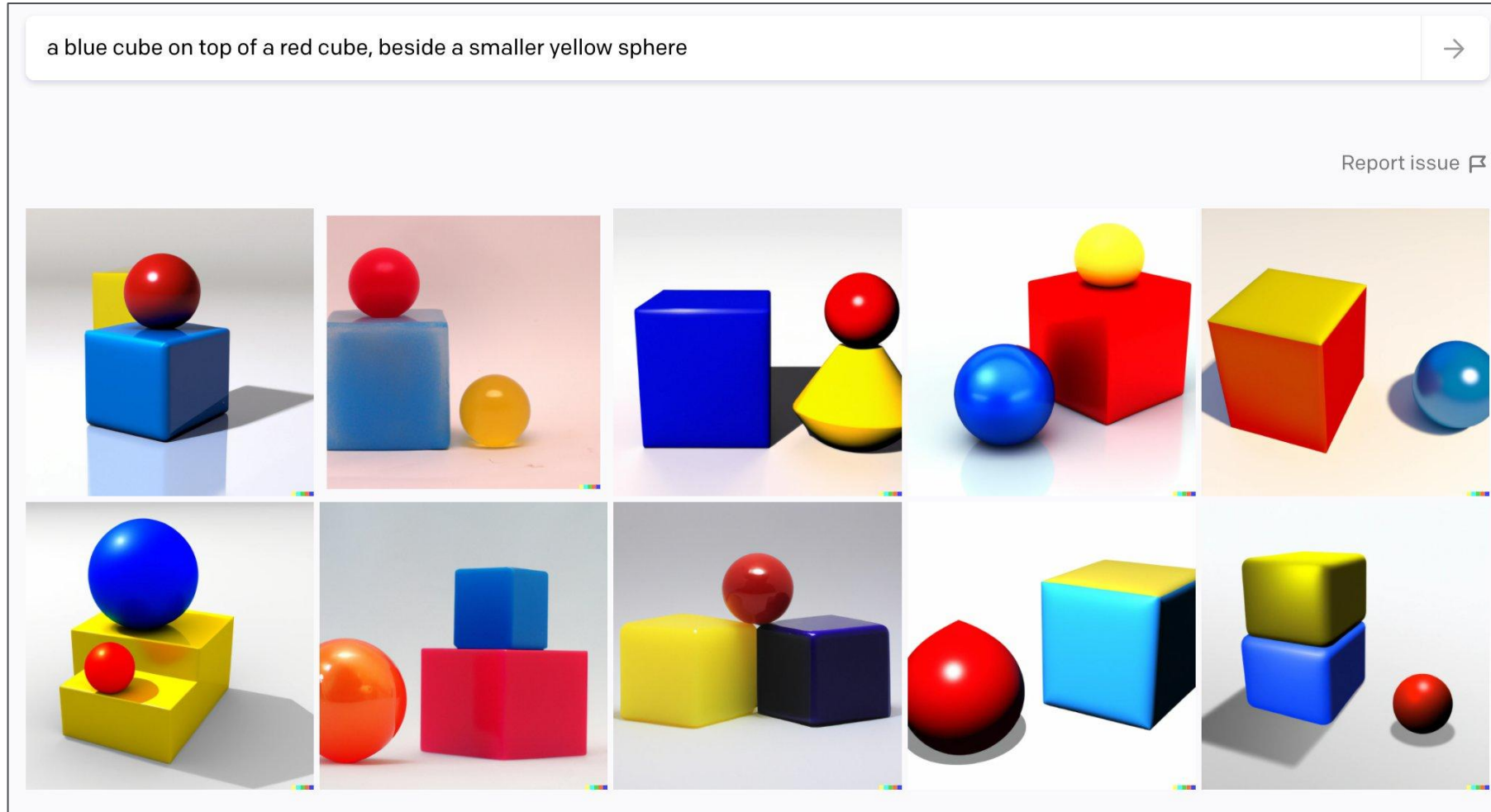


Adam × DALL-E

“a giant elephant losing it's
balance and falling into a
puddle in a bright sunny
desert”

Created with DALL-E, an AI system by OpenAI

OpenAI DALL-E 2



Imagine A Game With Infinite Adventures, As Unique As Your Own Life...

AI DUNGEON 2

Create your own world



Unlike virtually every other game in existence, you are not limited by the imagination of the developer in what you can do. Any thing you can express in language can be your action and the AI dungeon master will decide how the world responds to your actions.

Could a **purely self-supervised** Foundation Model
achieve grounded language understanding?

Self-supervision

1. The model's *only* objective is to learn co-occurrence patterns in the sequences it is trained on.
2. Alternatively: to assign high probability to attested sequences.
3. Generation then involves *sampling* from the model.
4. The sequences can contain anything.
5. The objective can't mention specific symbols or relations between symbols (no standard supervision).

Two paths to world-class AI chess?

Deeper Blue of the Future

1. Structured space of actions
2. Hard-coded rewards
3. Millions of games played

GPT-1000

1. Trained on billions of sequences of chess notation using only self-supervision:
You: Black [SEP] f4 d5 ... Qc7 [SEP]
White wins. [SEP]
You: Black [SEP] e4 e5 ... Qh3 [SEP]
Black wins. [SEP]
2. Bias in the training data for wins.
3. No separate notion of legal move, reward, etc.
4. When playing, simply generates new moves.

**Could a purely self-supervised Foundation Model
achieve grounded language understanding?**

Conceptions of semantics

David Lewis: “Semantics with no treatment of truth conditions is not semantics.”

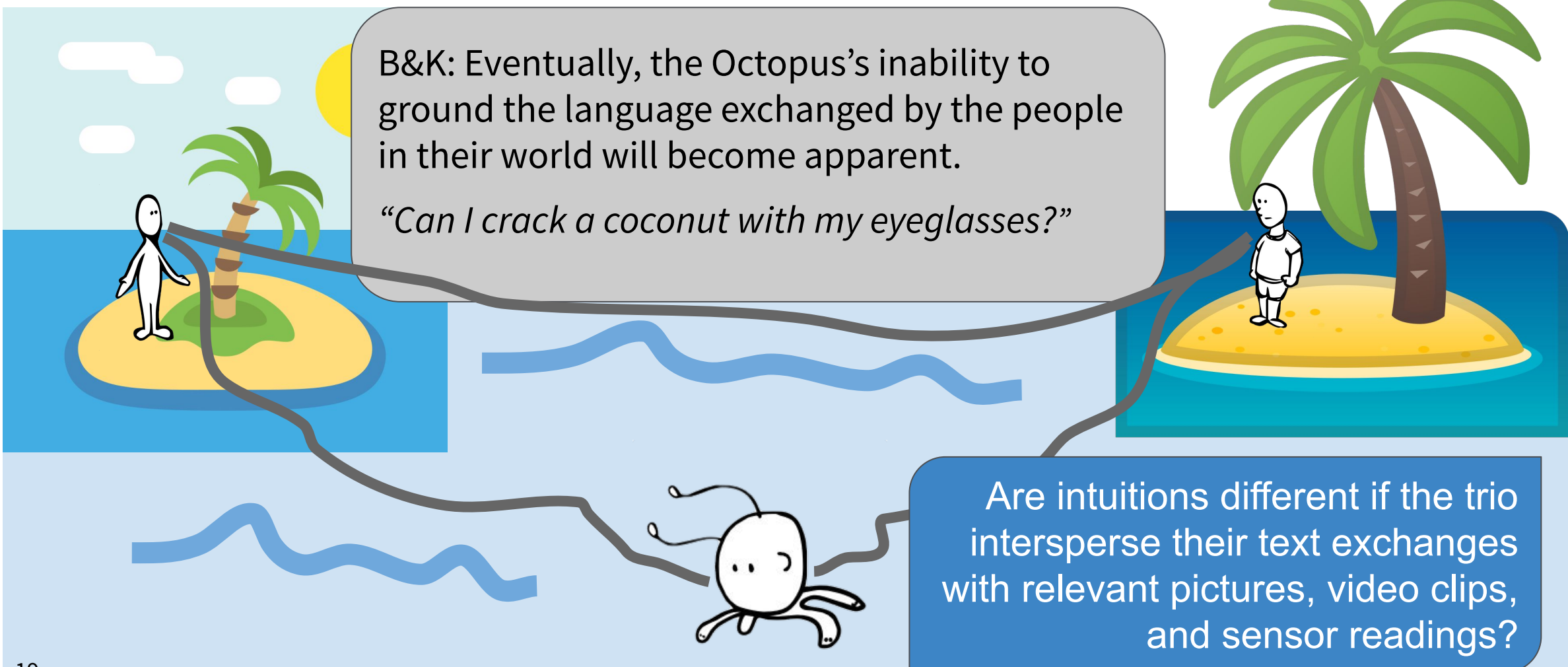
Jackendoff: Semantics as subjective, internalist judgment.

Jerrold Katz: “The arbitrariness of the distinction between form and matter reveals itself”

Natural logic: Language as proof system; model theory optional.

Will Merrill: Rich truth-conditional semantics can be induced from distributional data with certain biases towards discourse consistency.

Bender & Koller 2020: Symbol streams lack crucial information



B&K: Eventually, the Octopus's inability to ground the language exchanged by the people in their world will become apparent.

"Can I crack a coconut with my eyeglasses?"

Are intuitions different if the trio intersperse their text exchanges with relevant pictures, video clips, and sensor readings?

Multi-modal streams

Here's how you make a peanut  butter and jelly sandwich :




 \205\230\456\375\235\675 ...

Metaphysics and epistemology of understanding

Behavioral: understanding is purely dispositional and behavioral.

Internalism: understanding is achieving the right links between language and internal representations.

Referentialism: understanding is achieving the right links between language and the world.

Behavioral tests, once passed by AIs, are usually dismissed.

We need methods for *structural* analysis and assessment of models.

Behavioral testing: Tricky with Foundation Models

What is pragmatics?

Q: What is pragmatics?

A: The study of language use

What is semantics?

Q: What is phonology?

A: the study of systems of sounds in language

What do these fields have in common?

Q: What is semantics?

A:

[Questions about linguistics?]

Behavioral testing: Tricky with Foundation Models

Premise: every reptile danced

Hypothesis: every turtle moved

Label: entailment

Q: If every reptile danced, did every turtle move?

A: Yes.

Premise: no turtle ran

Hypothesis: a turtle moved

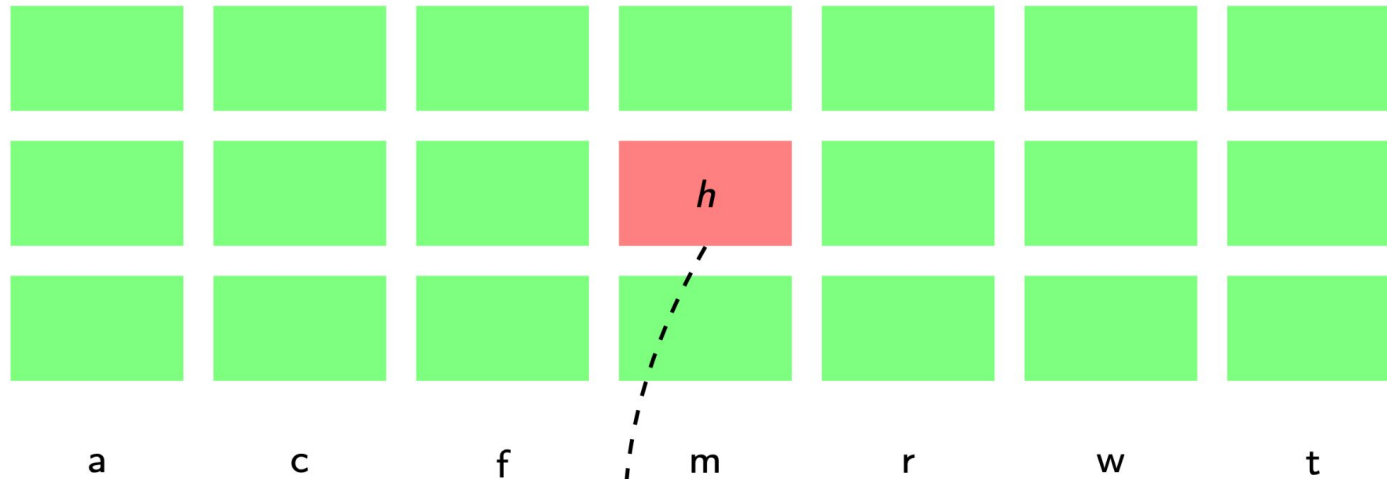
Label: [babbling]

Q: If no turtle ran, did a turtle move?

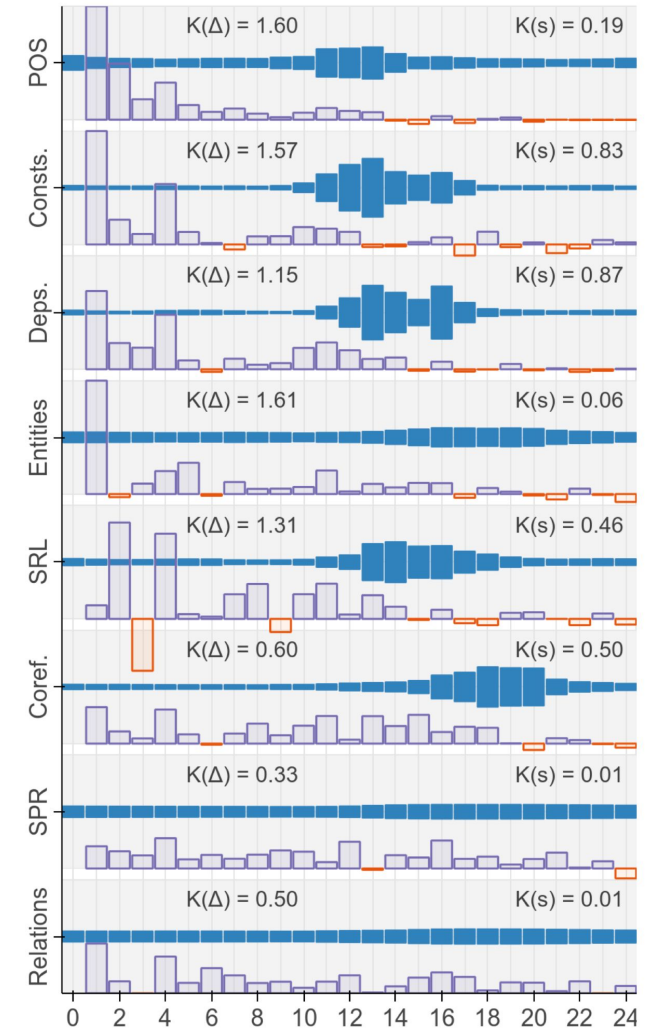
A: Maybe.

[Gao et al. 2020](#), [Shick and Schütze 2020](#)

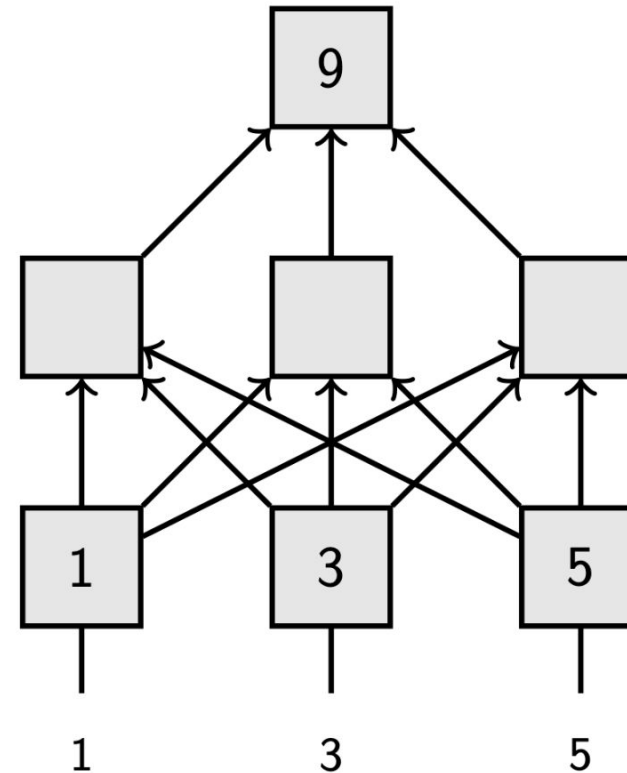
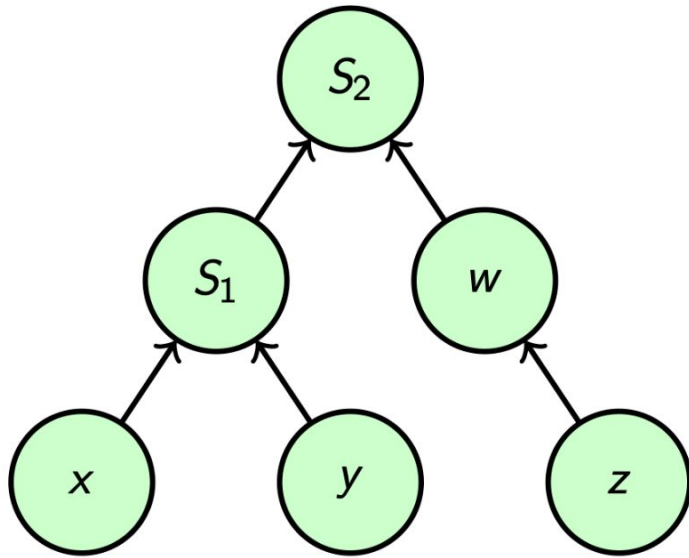
Internalism at work: Probing internal representations



$$\text{SmallLinearModel}(h) = \text{task}$$

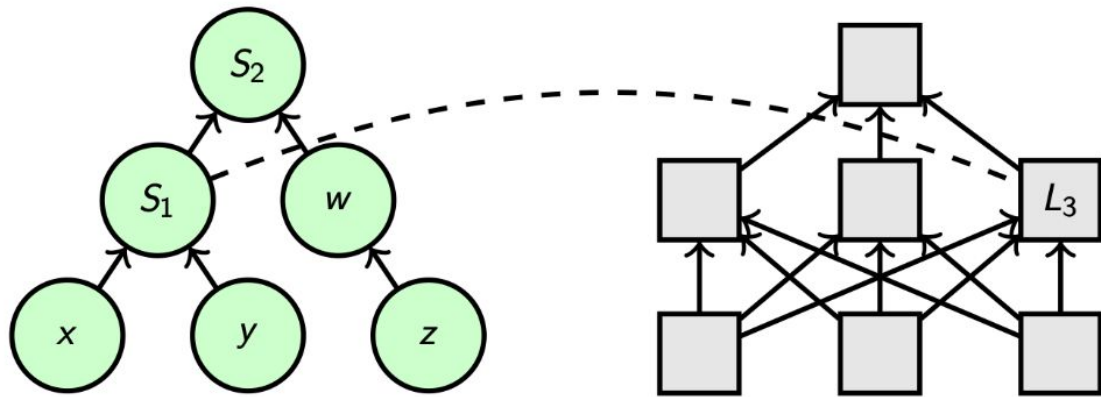


Internalism at work: Causal abstraction analysis

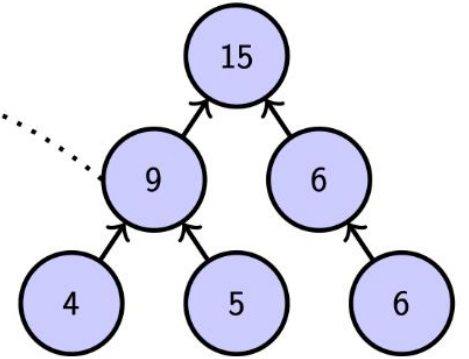
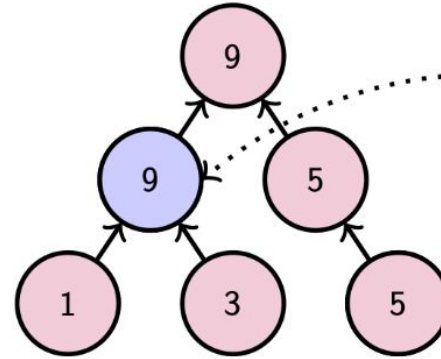
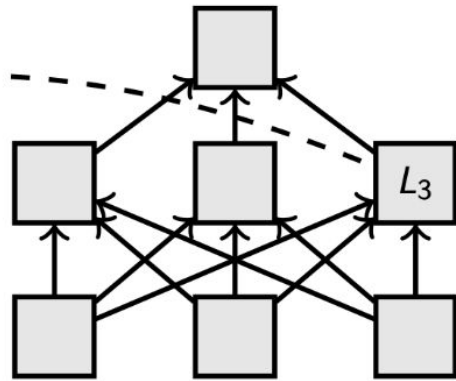
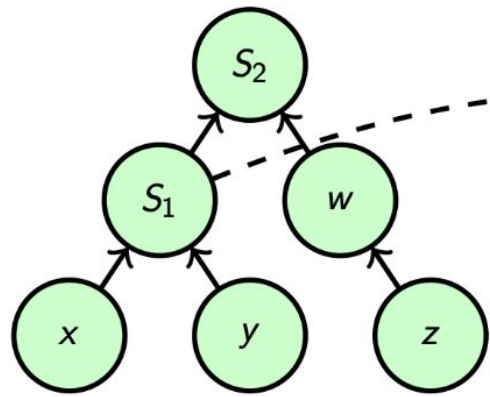


[Beckers et al. 2020](#); [Geiger, Lu et al. 2021](#); [Geiger, Wu, Lu et al. 2022](#); [Wu, Geiger et al. 2021](#)

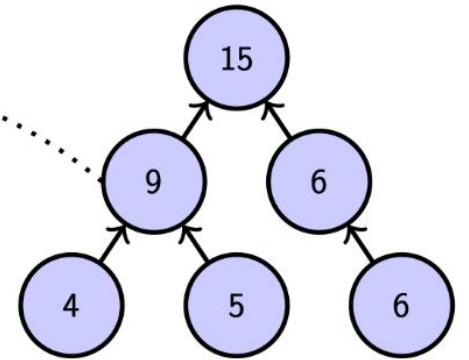
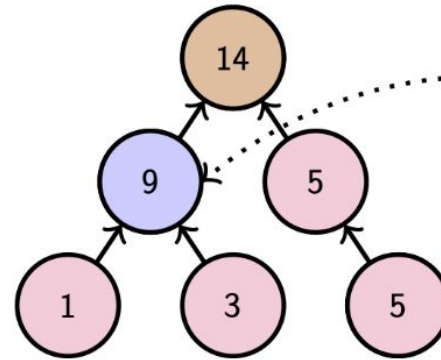
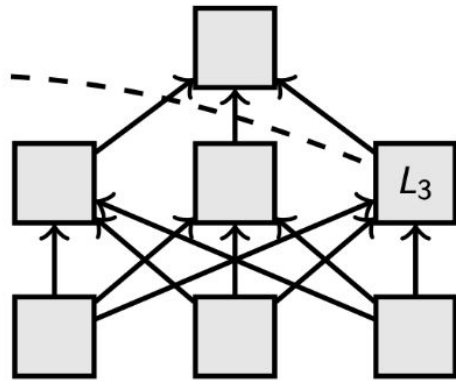
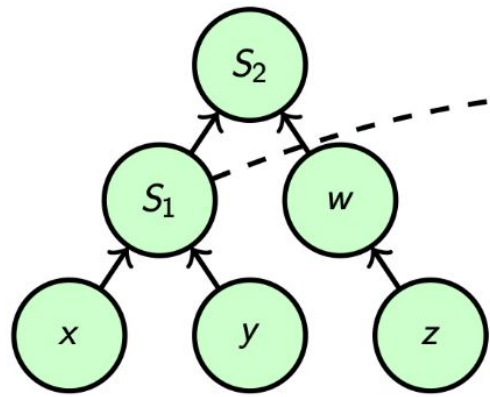
Internalism at work: Causal abstraction analysis



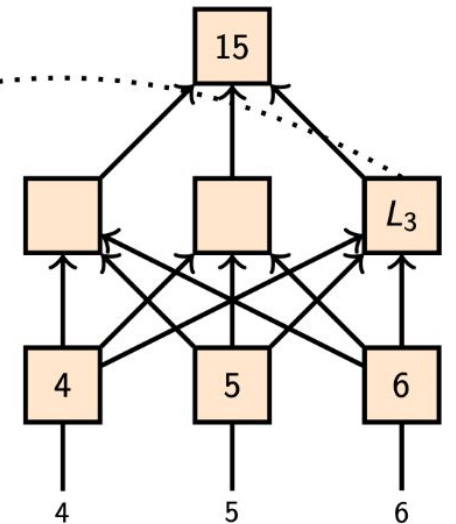
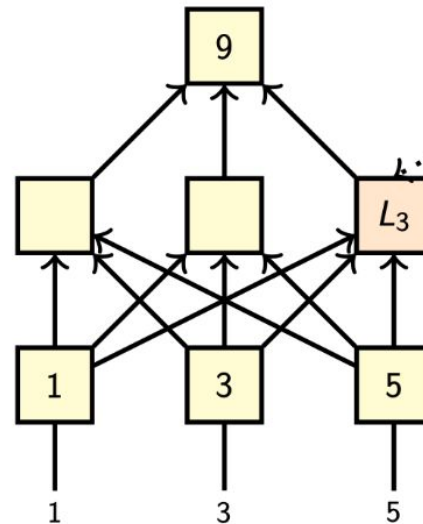
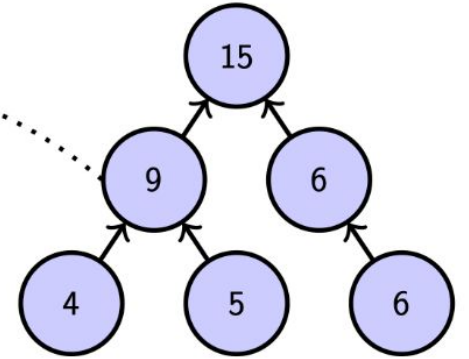
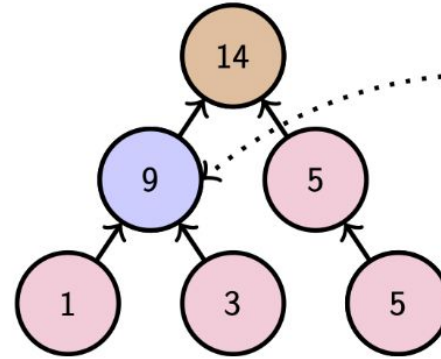
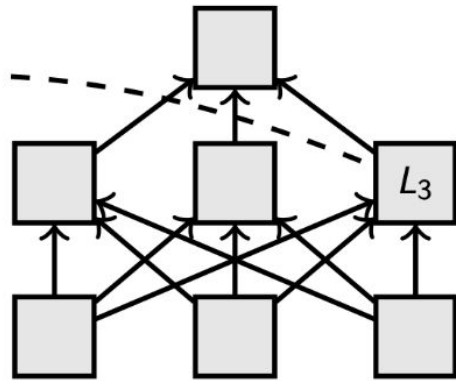
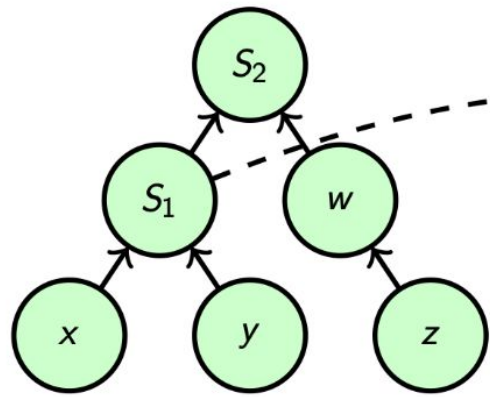
Internalism at work: Causal abstraction analysis



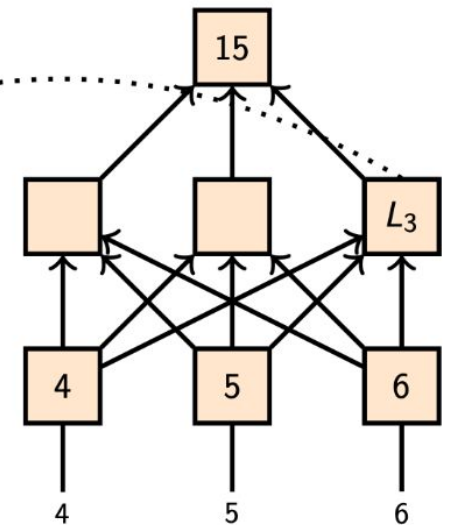
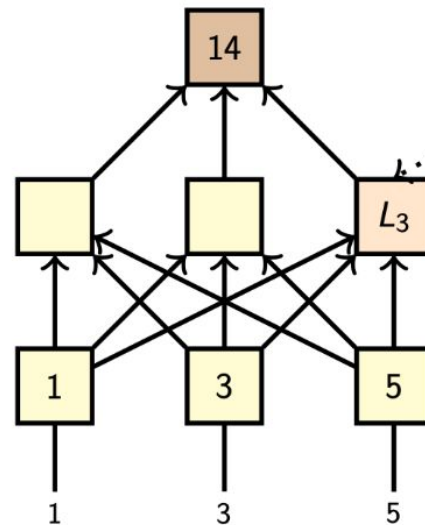
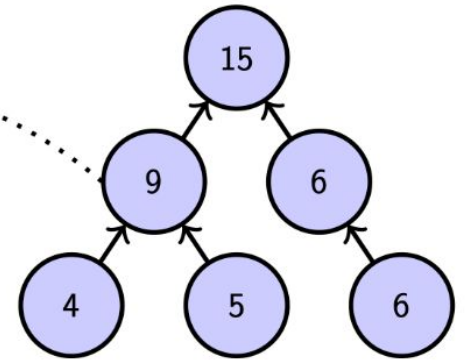
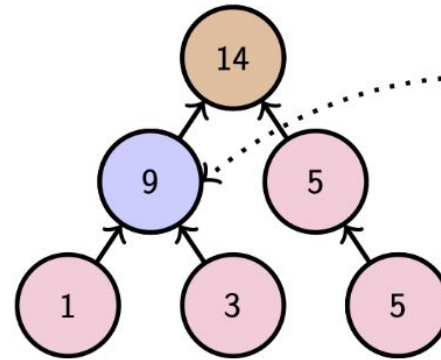
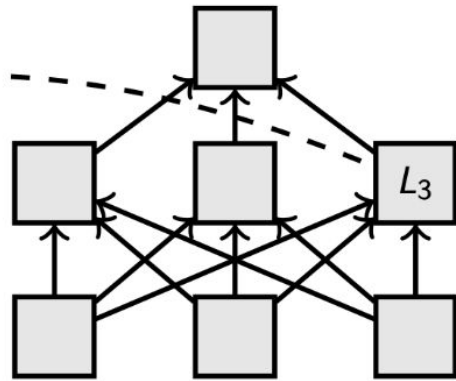
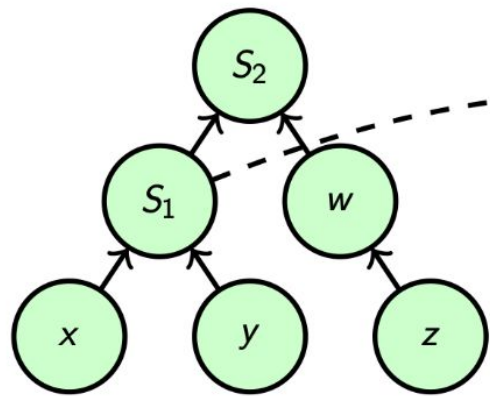
Internalism at work: Causal abstraction analysis



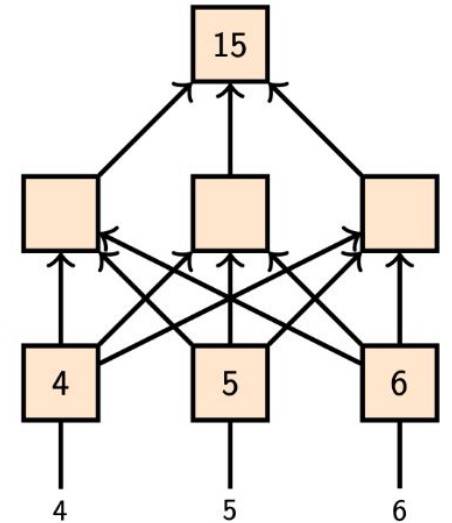
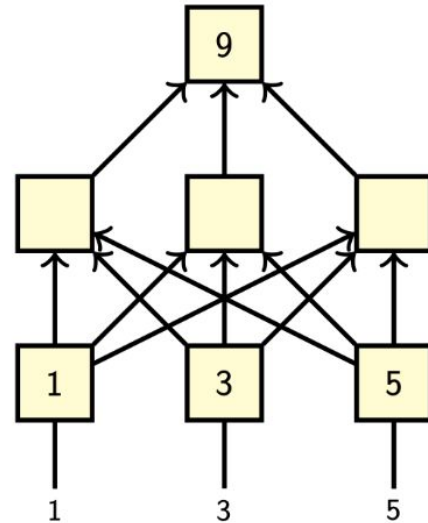
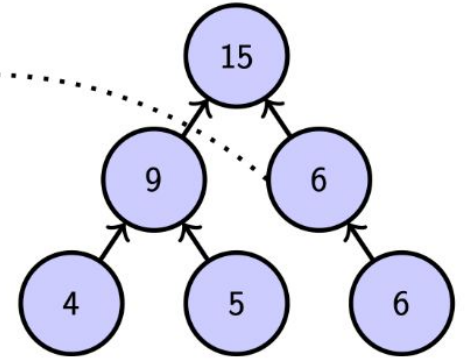
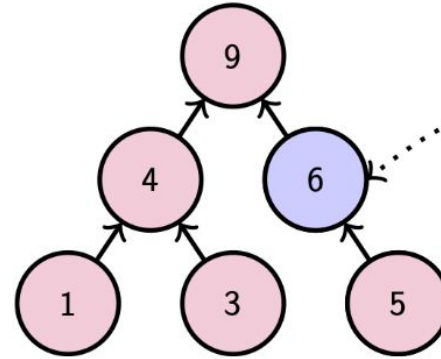
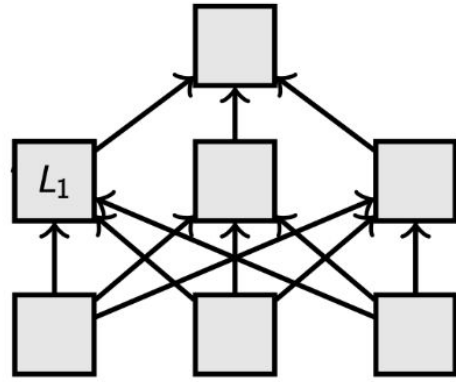
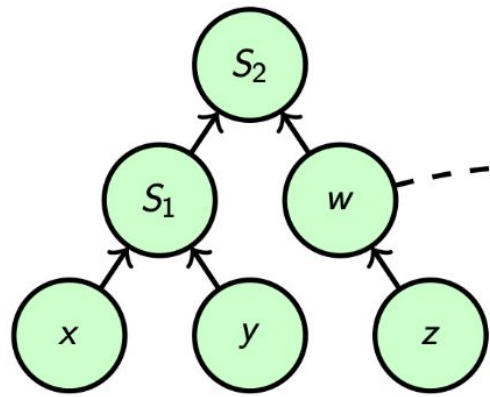
Internalism at work: Causal abstraction analysis



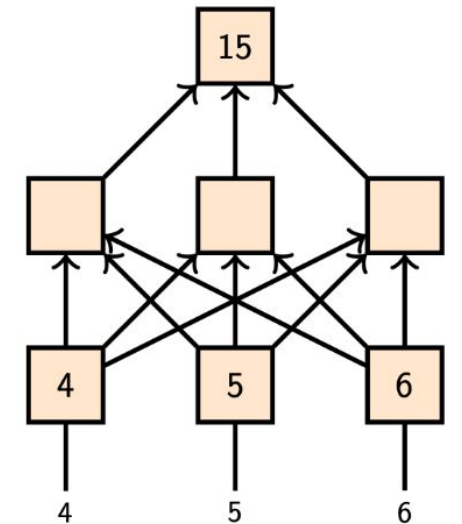
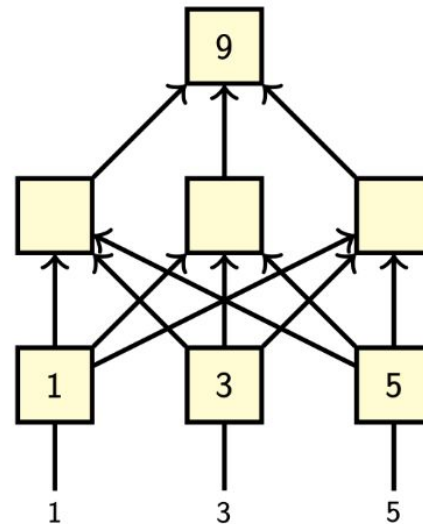
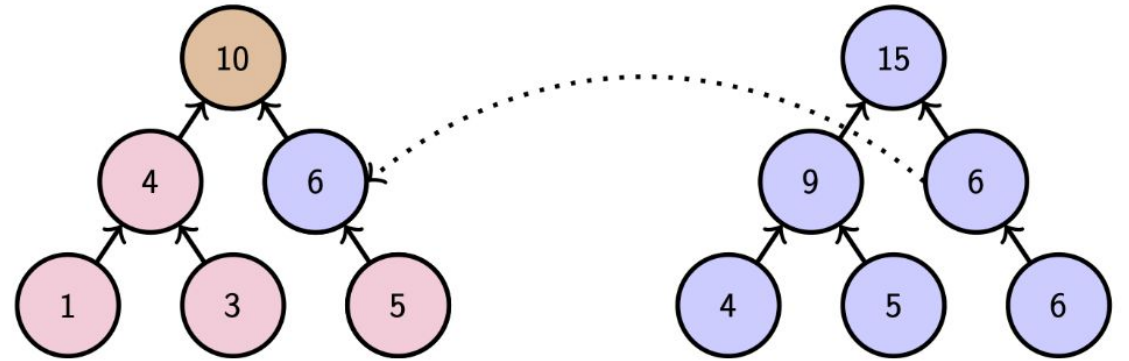
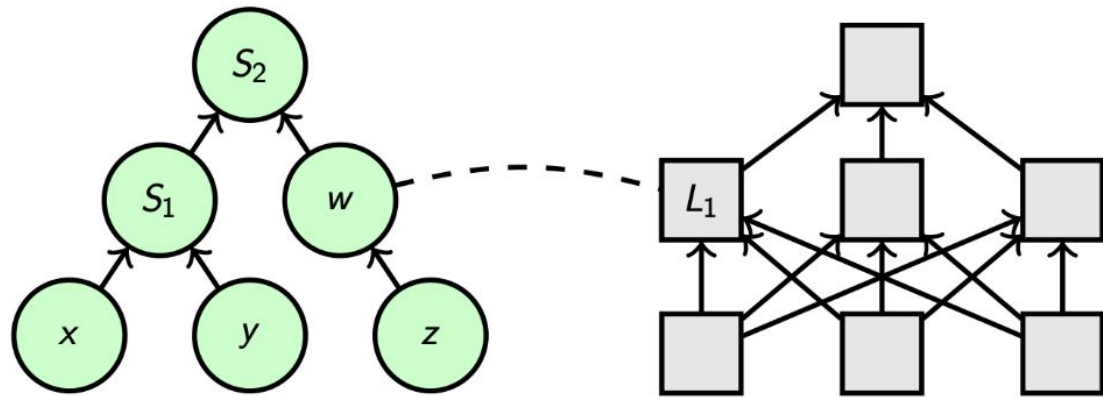
Internalism at work: Causal abstraction analysis



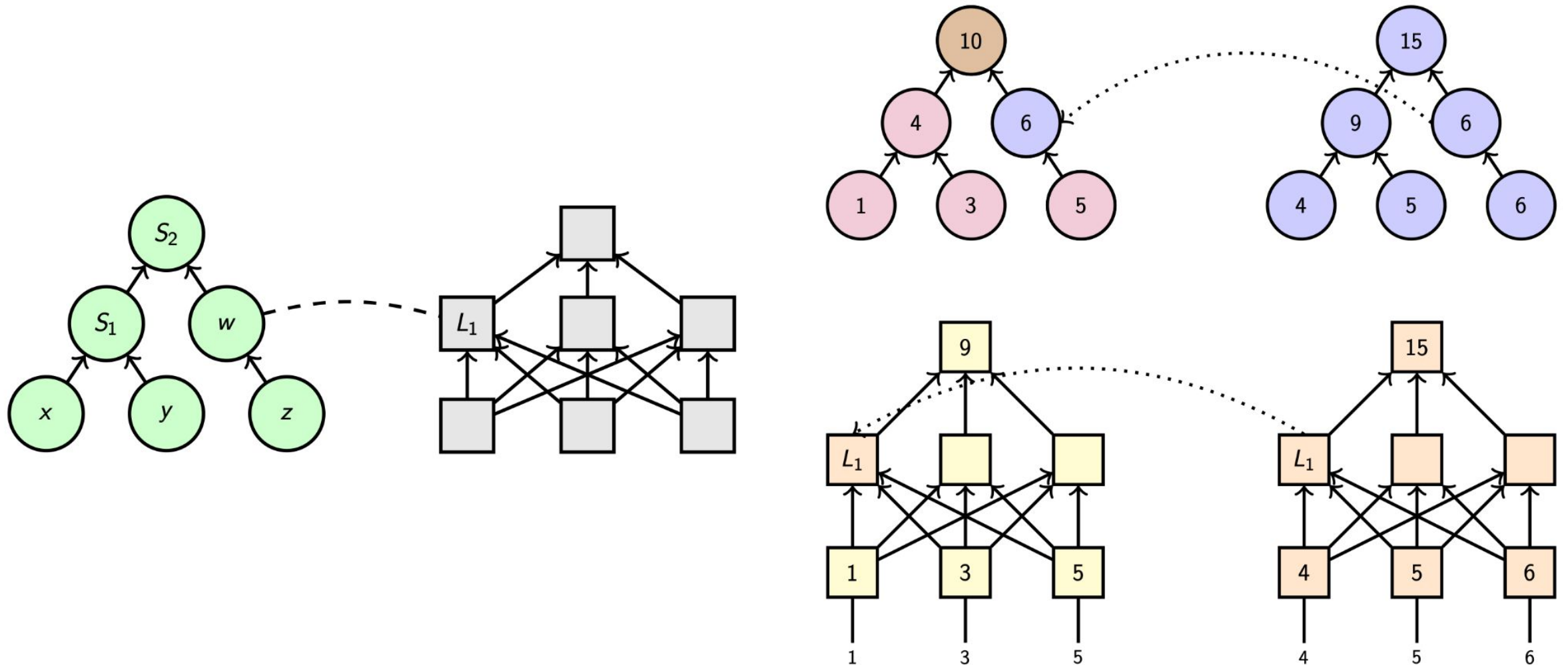
Internalism at work: Causal abstraction analysis



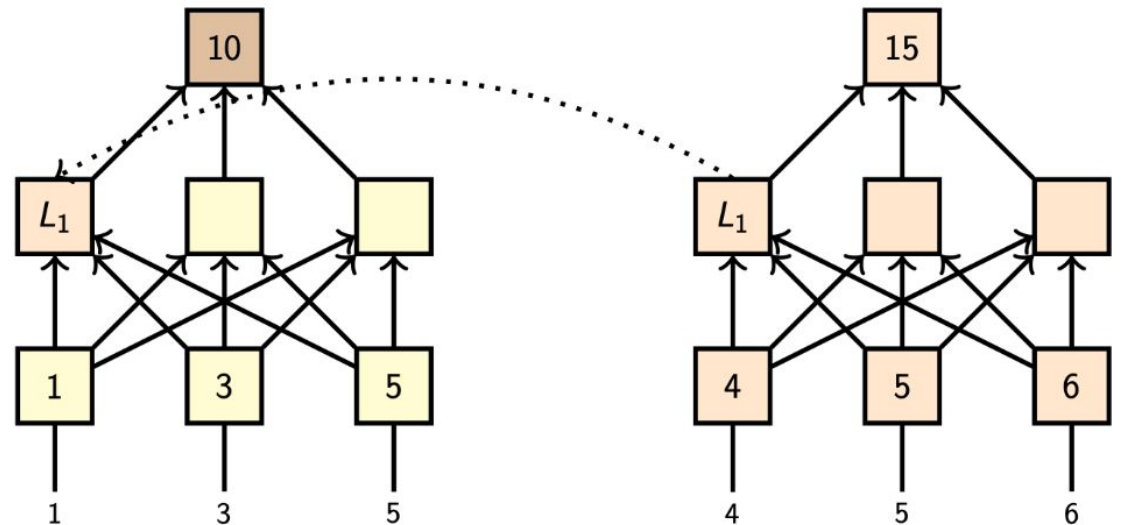
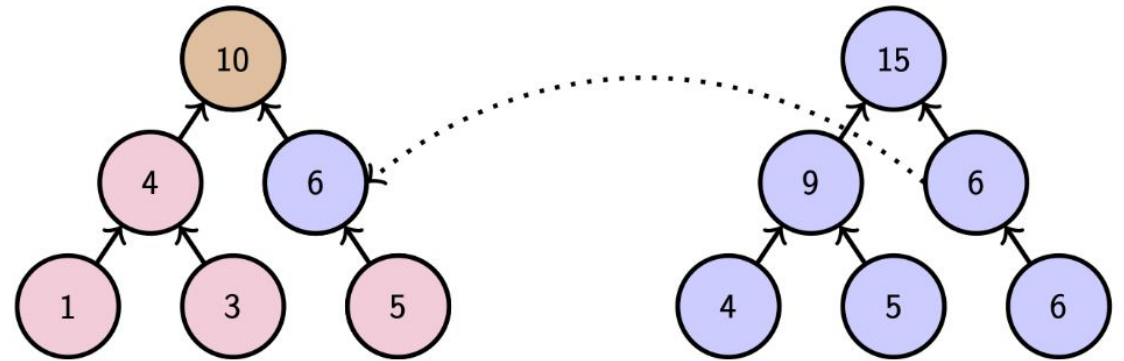
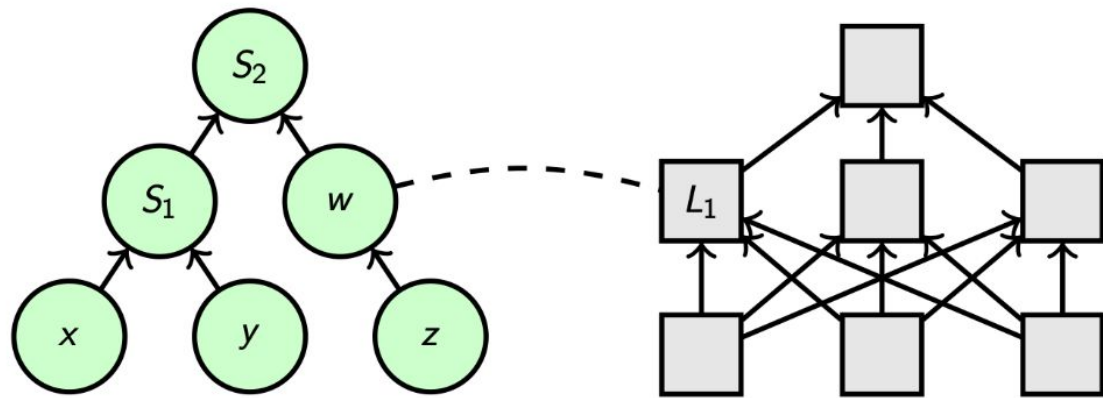
Internalism at work: Causal abstraction analysis



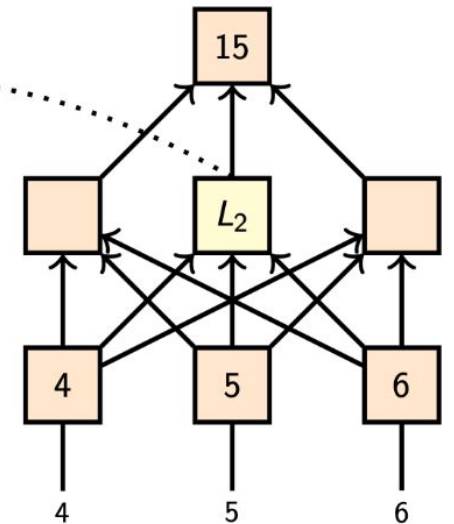
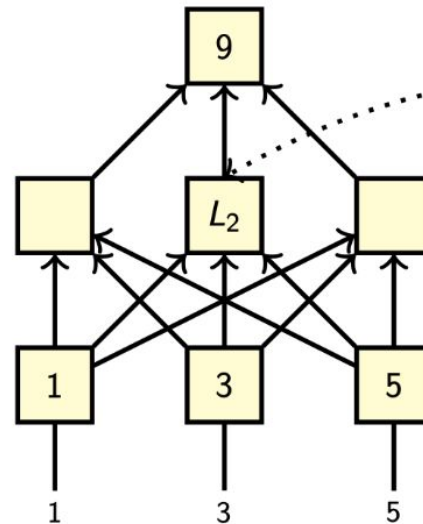
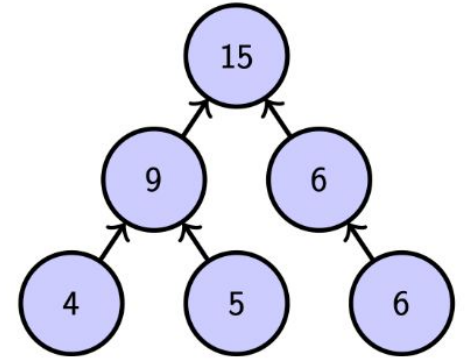
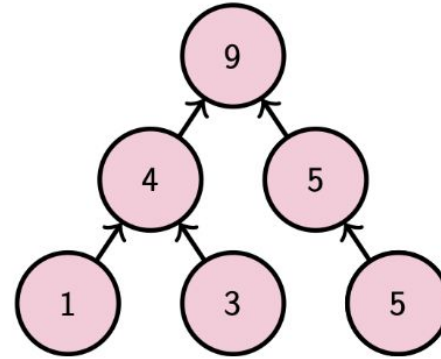
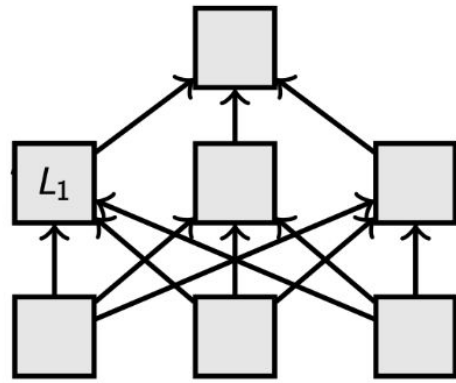
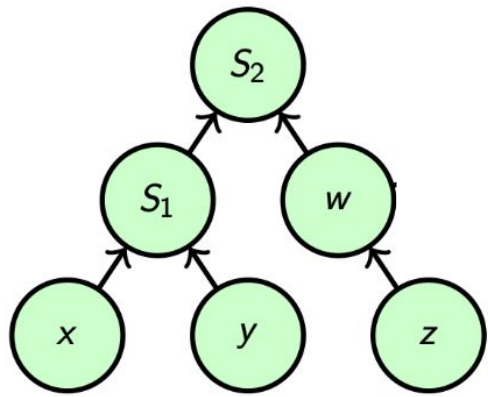
Internalism at work: Causal abstraction analysis



Internalism at work: Causal abstraction analysis



Internalism at work: Causal abstraction analysis



Findings of causal abstraction in large networks

- Fine-tuned BERT models succeed at hard, out-of-domain examples involving lexical entailment and negation *because* they are abstracted by simple monotonicity programs.
- Models succeed at the MNIST Pointer Value computer vision task *because* they are abstracted by simple programs like “if the digit is 6, then the label is in the lower left”.
- Models can be trained through interchange intervention training to better conform to high-level causal models/programs.

If a Foundation Model

- succeeds at hard language generalization tasks in a domain; and
- simulates a high-level causal model of that domain and the language used to describe it

then surely it has achieved grounded language understanding in that domain.

**Could a purely self-supervised Foundation Model
achieve grounded language understanding?**

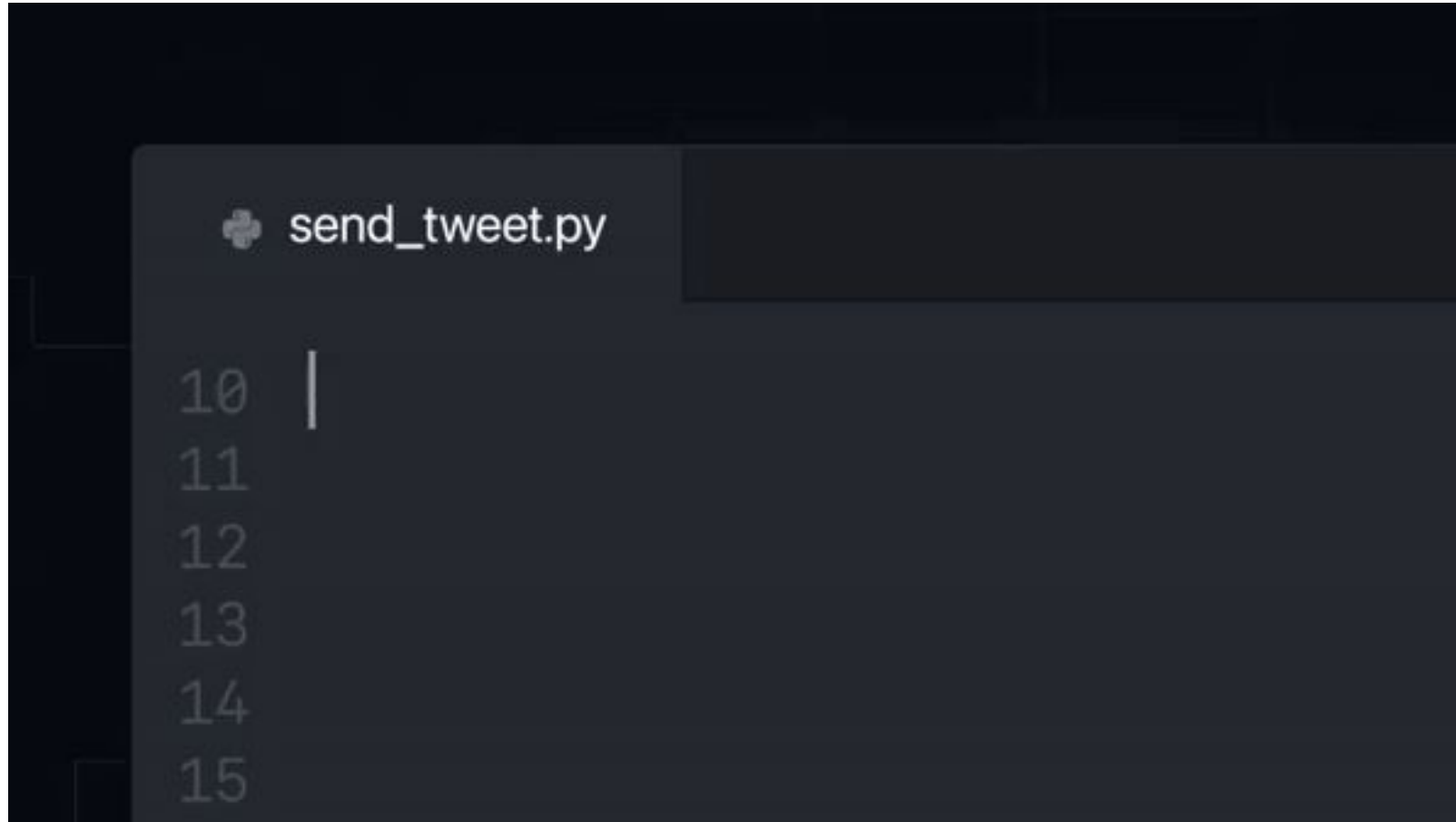
~~No (obviously not)~~

Yes (I don't see why not)

Thank you!

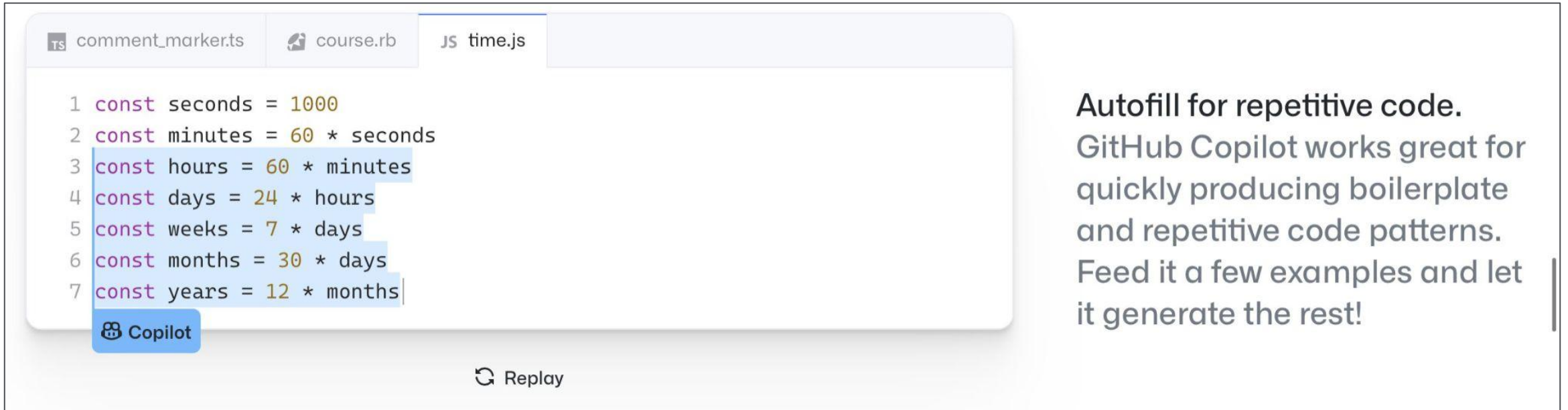
Appendix

Github Copilot (OpenAI Codex)



```
send_tweet.py  
  
10 |  
11  
12  
13  
14  
15
```


Github Copilot (OpenAI Codex)



The screenshot shows a code editor with three tabs: 'comment_marker.ts', 'course.rb', and 'JS time.js'. The 'JS time.js' tab is active and contains the following code:

```
1 const seconds = 1000
2 const minutes = 60 * seconds
3 const hours = 60 * minutes
4 const days = 24 * hours
5 const weeks = 7 * days
6 const months = 30 * days
7 const years = 12 * months
```

A blue Copilot suggestion box is visible at the bottom left of the code editor, containing the GitHub Copilot logo and the text 'Copilot'. Below the code editor, there is a 'Replay' button with a circular arrow icon.

Autofill for repetitive code.
GitHub Copilot works great for quickly producing boilerplate and repetitive code patterns. Feed it a few examples and let it generate the rest!

Pure self-supervision vs. regular supervision

Standard supervision for
nervous anticipation

My palms started to sweat as the lotto numbers were read off.	nervous anticipation = 1
I took a deep breath as the curtain started to rise on my debut night.	nervous anticipation = 1
I couldn't shake a deep feeling of unease about the whole affair.	nervous anticipation = 0

Foundation Model

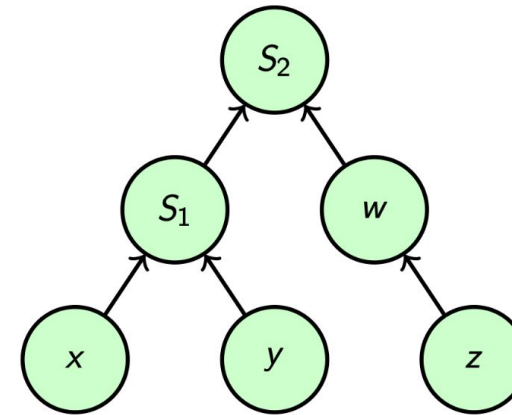
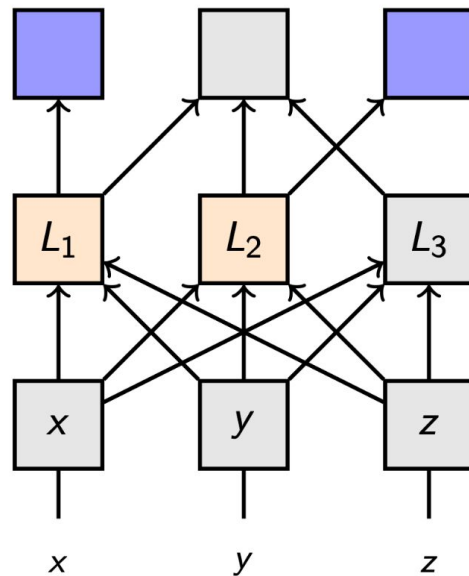
“Few-shot in-context learning”

Hey model, here is an example of nervous anticipation: “My palms started to sweat as the lotto numbers were read off.”

Hey model, here's an example without nervous anticipation: “...”

Probing does not support causal inferences

1. Probe L_1 : it computes z
2. Probe L_2 : it computes $x + y$
3. Aha!



4. But L_2 has no impact on the output!

$$W_1 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \quad W_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \quad W_3 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$$

$$\mathbf{w} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \quad (\mathbf{x}W_1; \mathbf{x}W_2; \mathbf{x}W_3) \mathbf{w}$$