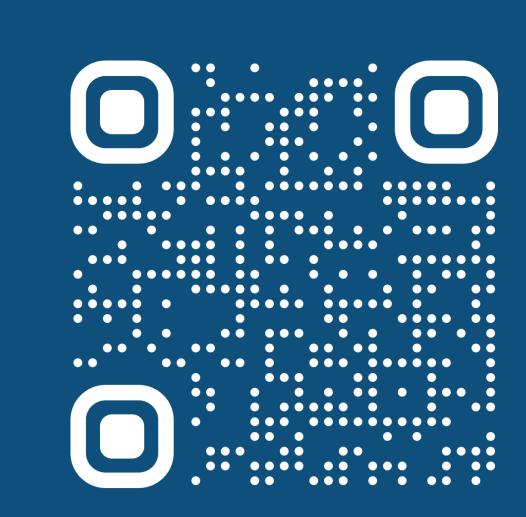


Brittle Minds, Fixable Activations: Understanding Beliefs Representations in Language Models

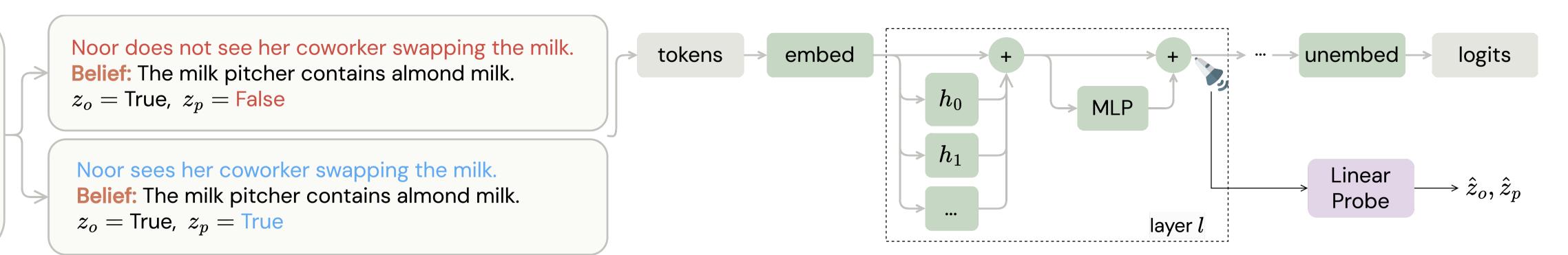


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Story: Noor is working as a barista at a busy coffee shop. Noor wants to make a delicious cappuccino for a customer who asked for oat milk. Noor grabs a milk pitcher and fills it with oat milk. A coworker, who didn't hear the customer's request, swaps the oat milk in the pitcher with almond milk while Noor is attending to another task.



Motivation

- Theory of Mind (ToM) is the ability to attribute mental states to oneself and others [1].
- Recent interest in evaluating Language
 Models' (LMs) generative performance on ToM tasks [2].
- Previous work suggests that LMs can represent beliefs of self and others [3].
- Experiments are limited in the number of models and settings studied, leaving several questions unanswered.

Research Questions

RQ1. Do internal belief representations emerge similarly in different LMs, and are they affected by model size and training?

RQ2. Are internal belief representations structured or the results of spurious correlations?

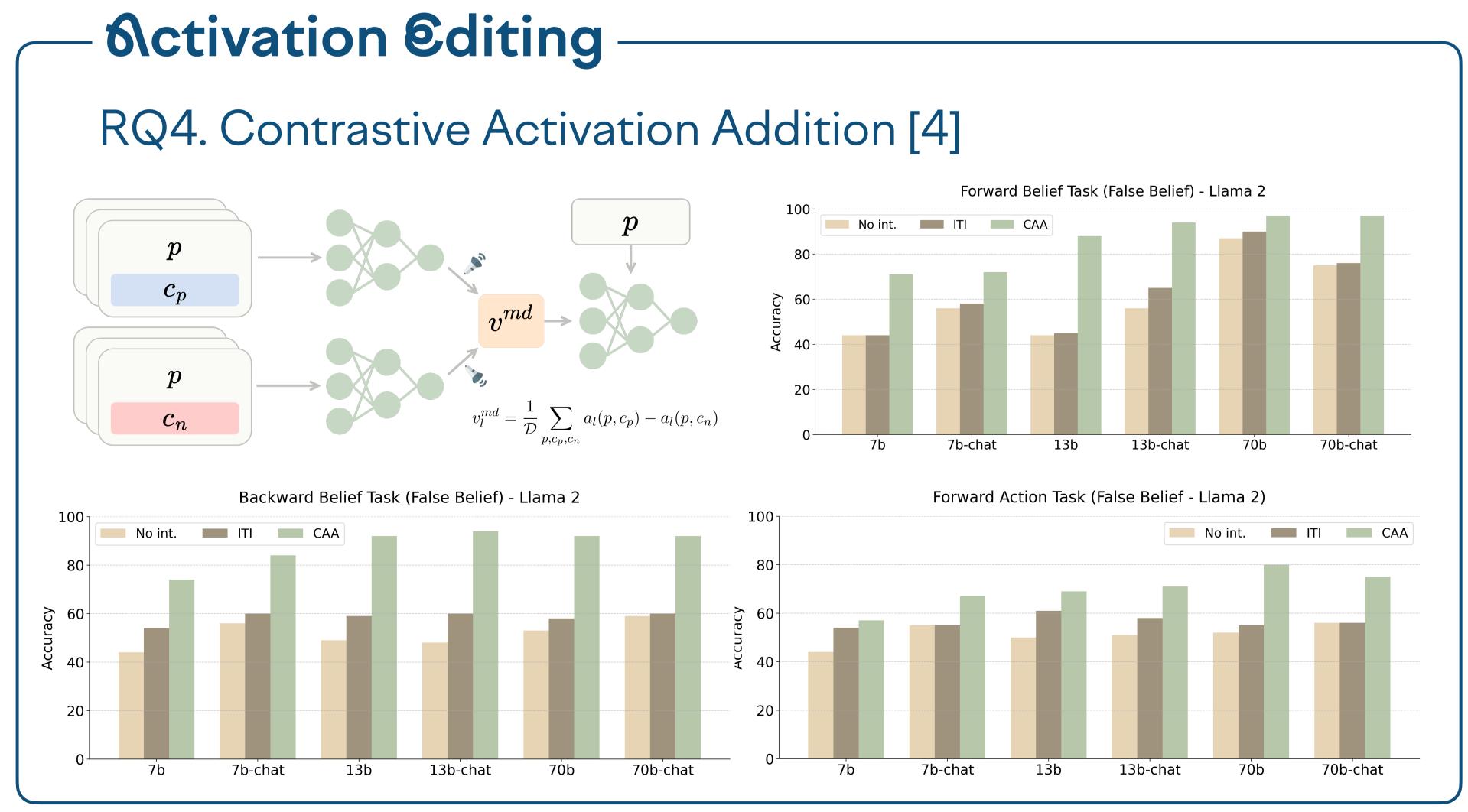
RQ3. Are internal belief representations robust?

RQ4. Can we enhance LMs' performance by editing their activations without training dedicated probes?

Contributions & Findings

- 1. Extensive probing experiments across **12 models**, suggesting that:
 - The representations of others' beliefs of others improve with model size and, more crucially for smaller models, fine-tuning.
 - Probes capture structured belief representations rather than spurious correlations.
 - LMs' representations of others' beliefs are brittle to prompt variations.
- 2. We show that by using contrastive activation addition [4] it is possible to improve models' ToM performance by steering their activations without the need to train any probe.

Probing Language Models' Representations RQ1. Effect of size and fine-tuning Llama-2, oracle Llama-2, protagonist Pythia, oracle Pythia, protagonist 60 Model depth Model depth Model depth Model depth RQ2. Control tasks Llama-2-13B Llama-2-70B Llama-2-7B Llama-2-7B-chat Llama-2-13B-chat Llama-2-70B-chat 20 20 50 Llama-2-7B Llama-2-70B-chat Llama-2-7B-chat Llama-2-13B Llama-2-13B-chat Llama-2-70B Layer Layer Layer Layer Layer Layer RQ3. Robustness to prompting Llama-2-13B Llama-2-7B Llama-2-7B-chat Llama-2-13B-chat Llama-2-70B Llama-2-70B-chat Layer Layer Layer Layer Layer Layer



References