

Rethinking Reasoning: When Next-Token Prediction Mimics Thought

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Background

Dual-System Thinking

System 1	System 2
Fast, Direct Answer	Deliberate, Reasoning

“System 2” Thinking in Large Language Models

LLMs simply predict the next token (as system 1), yet with recent techniques they can produce *multi-step reasoning* (as system 2),.

Research Question

How does reasoning-like behavior arise from the basic predictive mechanism — instead of a dedicated reasoning module?

Methods

Task

➤ Multi-step arithmetic problems (e.g., $12 + 7 - 19 = ?$)

➤ **Base Model:** GPT-Neo-1.3B

- Not trained for reasoning;
- Poor at multi-step arithmetic;
- Good enough for one-step arithmetic.

Post-Training Setup

Fine-tuned GPT-Neo multiple times with different proportions of A & B.

Type	Example
A: Immediate Answer	$19 - 8 - 2 = ?$ [answer] = 9
B: Step-by-Step Reasoning	$13 + 19 - 6 = ?$ [thinking] = $32 - 6$ [answer] = 26

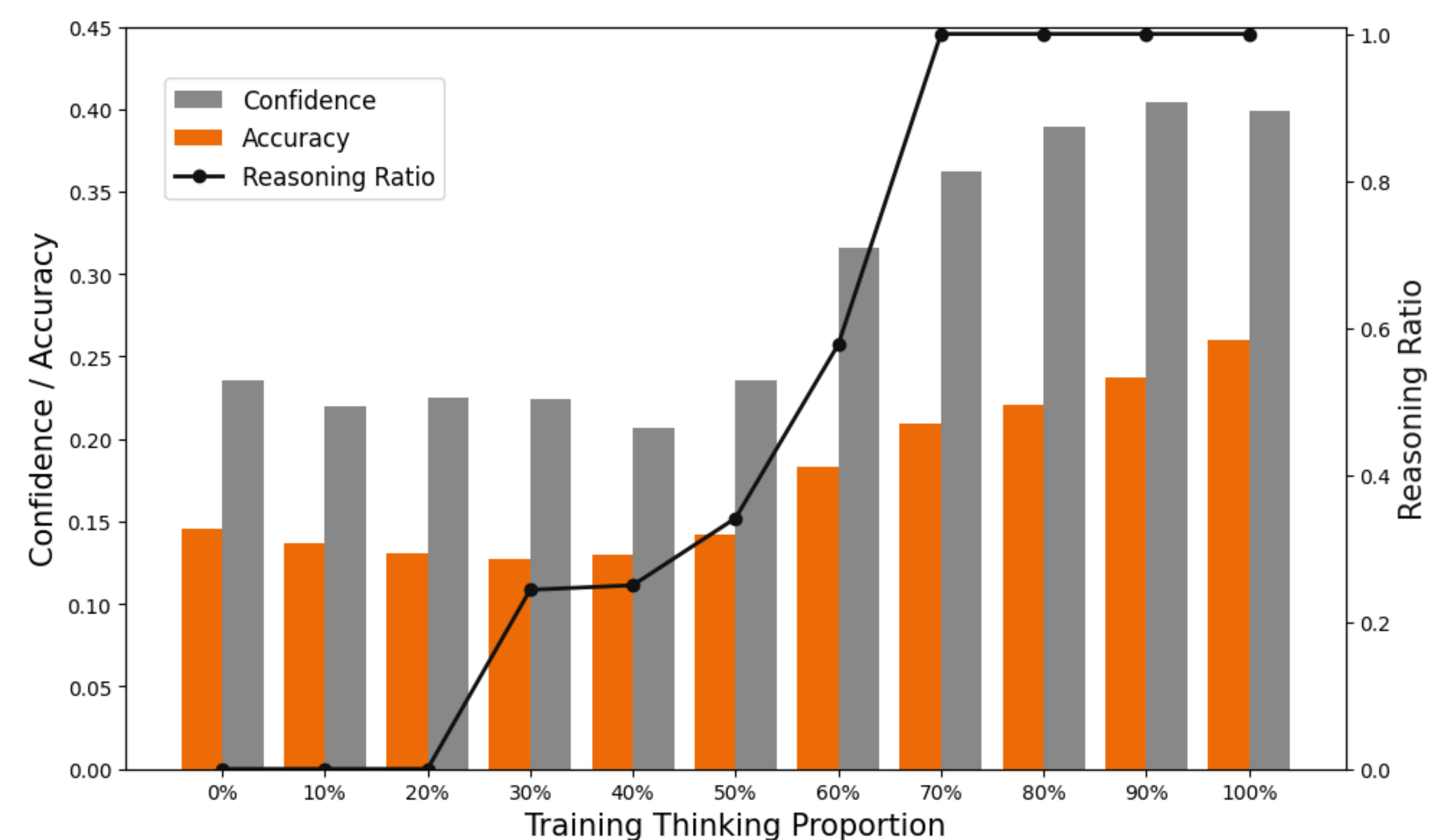
(e.g., 90% A & 10% B)

Evaluation:

- Test on **unseen problems** (e.g., “ $12 + 7 - 19 = ?$ ”)
- Test on **unseen problems with reasoning cue** (e.g., “ $12 + 7 - 19 = ?$ [thinking]”)

Results

- Immediate-answer training (100% A & 0% B) → almost no intermediate reasoning;
- **More reasoning training samples** (>50% B) → More reasoning output, with higher accuracy and confidence;



- Even with **minimal reasoning training** (90% A & 10% B), prompting with “[thinking]” could trigger reasoning and improved performance.

Discussion

“Reasoning” is not a new capacity added onto LLMs , but an outcome of **reshaping the model’s predictive pathways**.

- Early LLMs default to shortcut-answer pathway because **most training data emphasizes direct answers**.
- “Reasoning” techniques made *intermediate-step* tokens more frequent or more available, then “reasoning” becomes the new optimal generation.

Engineering Insight:

Future LLM “reasoning” methods can **explicitly boost the likelihood of reasoning-related tokens** to encourage multi-step outputs.

Cognitive Insight:

- “System 2” may not be a distinct mental faculty.
- It could also be the **suppression of fast, intuitive (“System 1”) responses**, allowing thought to unfold through intermediate steps before a final answer.

Take-away

In both LLMs and humans, reasoning may emerge from short-term prediction—**shaped by experience (training) or activated by context (prompt)** to favor multi-step trajectories over immediate answers.