

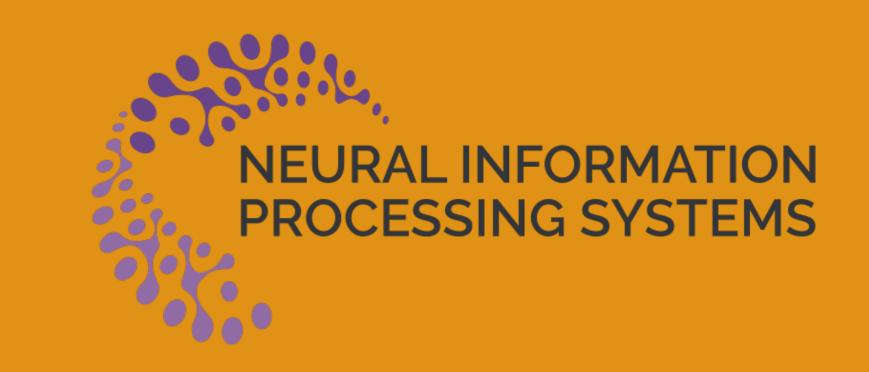


Google DeepMind

Topology of Reasoning

Understanding Reasoning from Graph-Theoretic Lens

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Motivation

- Reasoning models (e.g., o3) achieve success

 They solve expert-level coding, math, and science problems by thinking.
- How do they reason? How do they differ?

The mechanisms behind the reasoning abilities remain poorly understood.

Our lens: Reasoning Graphs

By modeling internal hidden states as structured graphs, we uncover what distinguishes reasoning models at the representational level.

Question

Joy can read 8 pages of a book in 20 minutes. How many hours will it take her to read 120 pages?

Generated Answer Base Model

In 1 hour, there are 3 sets of 20 minutes.

8 pages ÷ 20 min=0.4 pages/minute.

120 pages ÷ 0.4 pages/minute=360 minutes

360 minutes ÷ 60 minutes/hour=6 hours.

The answer is 6 hours.

Incorrect

Reasoning Model

In 1 hour, there are 3 sets of 20 minutes.

8 pages ÷ 20 min=0.4 pages/minute.

120 pages ÷ 0.4 pages/minute=360 minutes.

360 minutes ÷ 60 minutes/hour=6 hours.

Wait, let me verify the calculations. ...

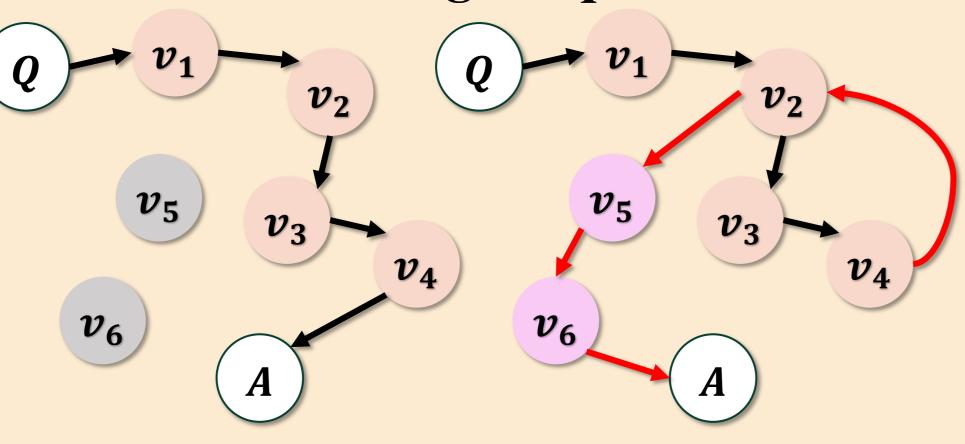
8 pages ÷ 20 min=0.4 pages/minute.

120 pages ÷ 0.4 pages/minute=300 minutes.

300 minutes ÷ 60 minutes/hour=5 hours.

The answer is 5 hours.

Reasoning Graph



Reasoning Graph from Internal State

■ Segment Representation

For each question, reasoning steps: $R = (r_1, r_2, ..., r_T)$ Segment representation by averaging $h_i^{\ell}(t)$: $s_i^{\ell} = \frac{1}{L} \sum_{t=1}^{L_i} h_i^{\ell}(t)$

Node Definition

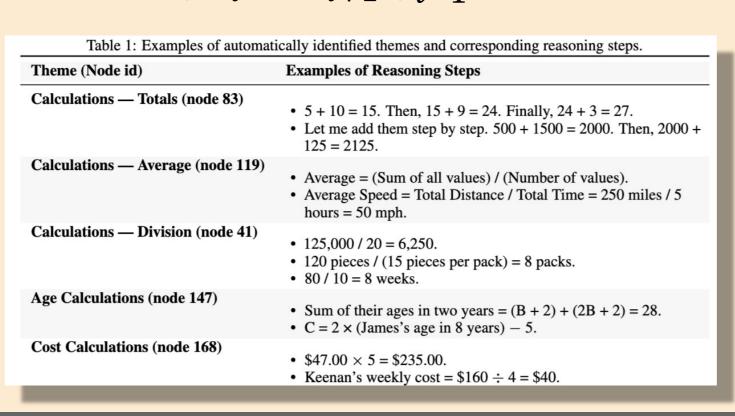
Segment s_i^{ℓ} clustered using K-means. Cluster center c_k becomes a graph node v_k :

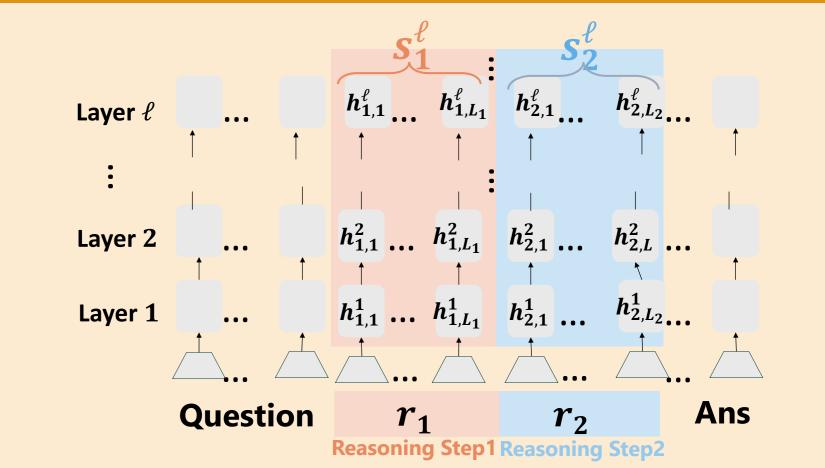
 $V = \{v_1, \dots, v_K\}, \qquad d(v_i, v_j) = ||c_i - c_j||^2$

Edge Construction

For each question, connect nodes in order.

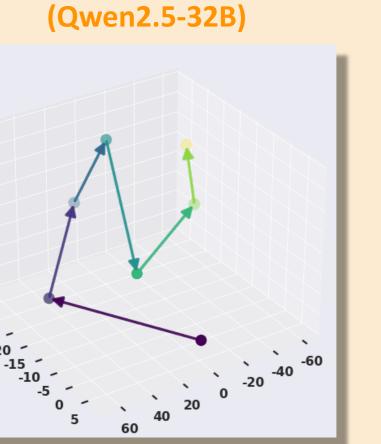
$$E = \{(v_i \to v_{i+1})\}_{i=1}^{T-1}$$

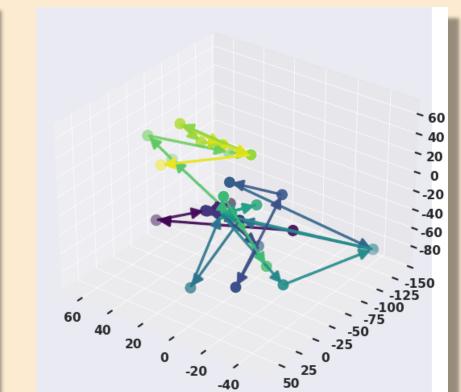




Base Model (Qwen2.5-32B)

Reasoning Model (DeepSeek-R1-Qwen-32B)





Key Graph Properties

Reasoning Model

Cycles

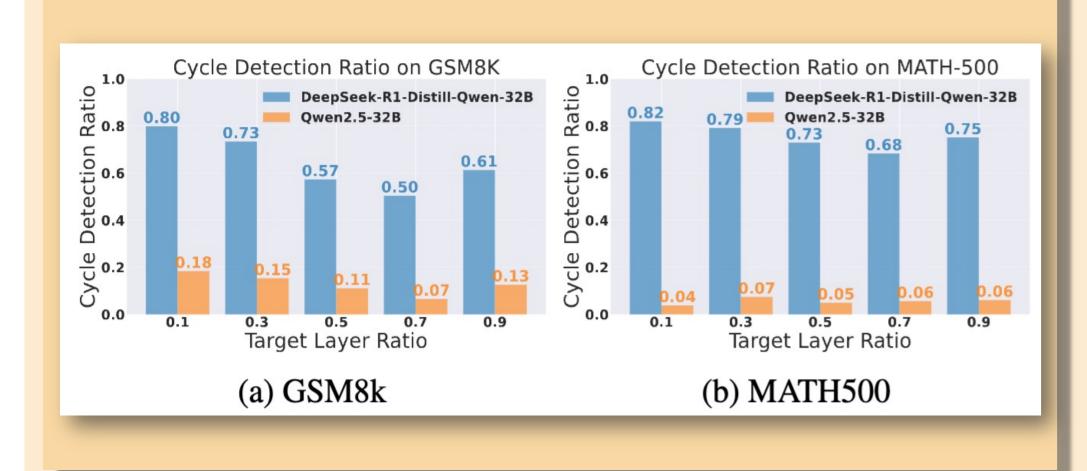
Measurement

Detect cycles in the reasoning graph, defined as repeated visits to the same node, excluding. self-loops.

Result

Reasoning models have *more cycles*.

Difficult tasks induce more cycles



Diameters

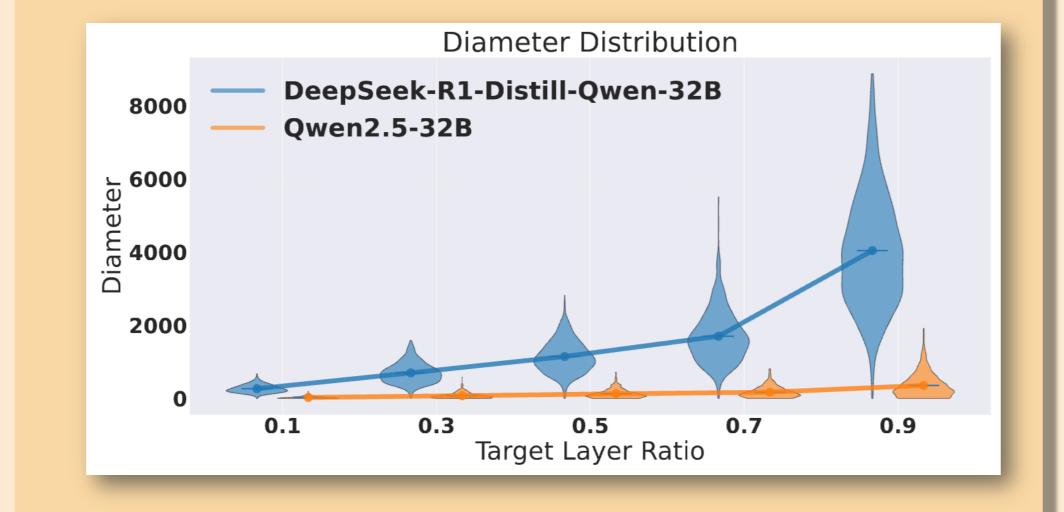
Measurement

Maximum shortest path distance between any two reachable node Diameter = $\max_{u} \max_{u \neq v} d(u, v)$

Result

Base Model

Reasoning models *larger diameters*They explore broader spaces



Small-World

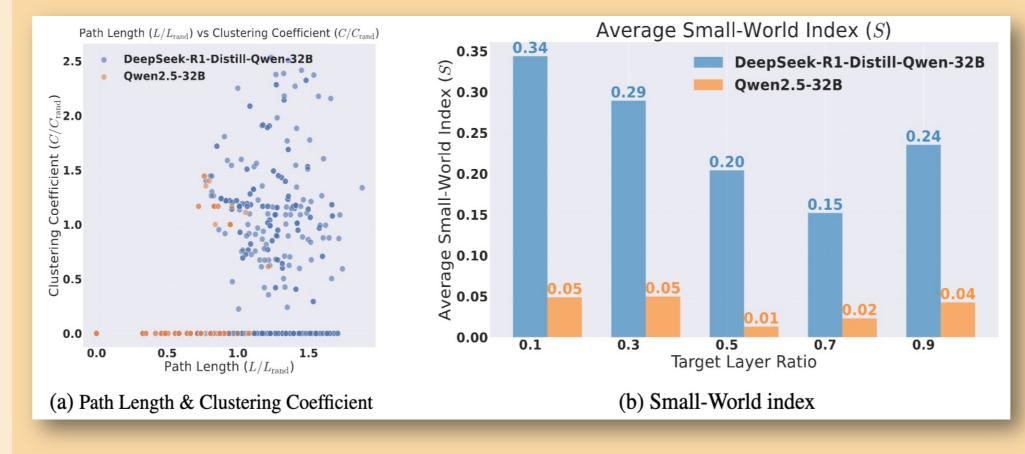
Measurement

Common in networks—SNS, biology, etc.
Clustering coefficient / Path length.

Smallworld =
$$\frac{C/C_{rand}}{L/L_{rand}}$$

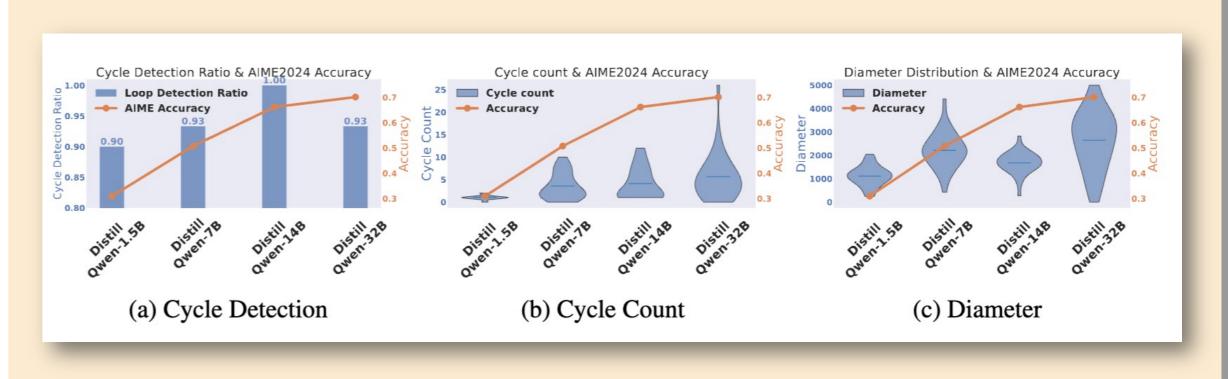
Result

Reasoning models show higher *C* and longer *L* forming small-world graphs



Model Size Matters

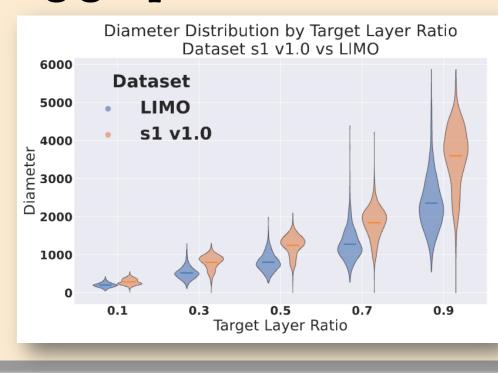
Larger models show more cycles, broader graphs. As model size increase (1.5B \rightarrow 32B), reasoning graph exhibit more cycles and larger diameters.



Practical Takeaway: SFT

Better SFT data = Better reasoning graphs sl prompts yield larger reasoning graphs than LIMO.

Graph properties could help evaluate and build better reasoning-SFT data



[1] Niklas et. al., "s1: Simple test-time scaling", [2] Ye et. al., "LIMO: Less is More for Reasoning"