

Speculative Monte-Carlo Tree Search

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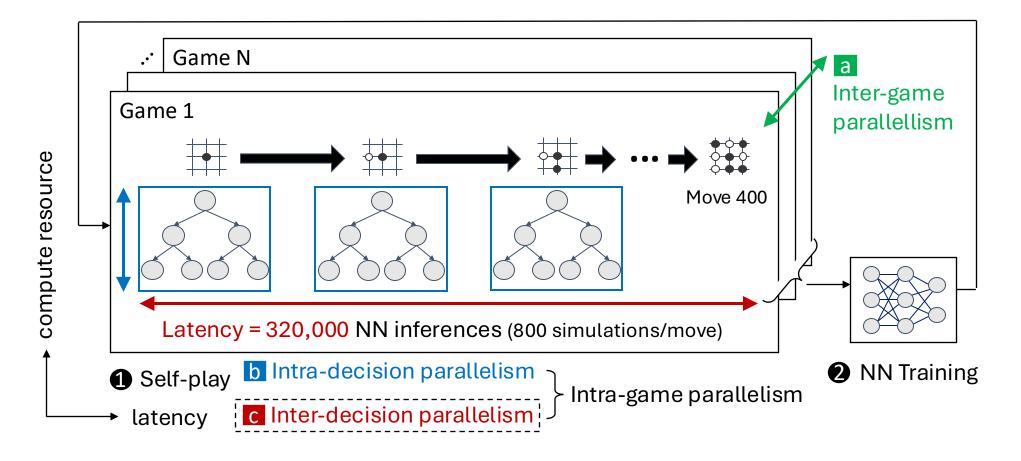
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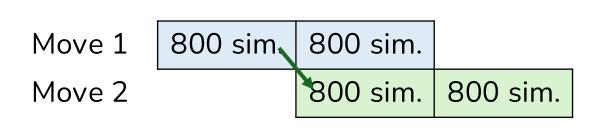
Sequential MCTS limits AlphaZero training latency

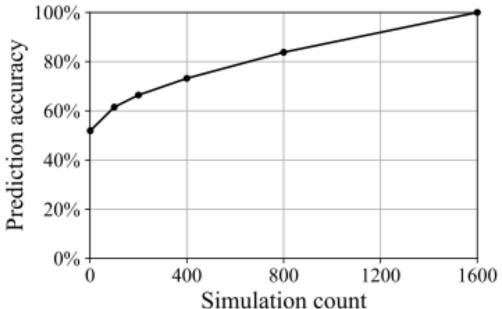
AlphaZero has achieved breakthroughs in chemistry, medicine, and many other fields, but the training is considerably time-consuming.



Speculative Parallelism in MCTS

Motivation: partial MCTS can reasonably predict the full MCTS results.





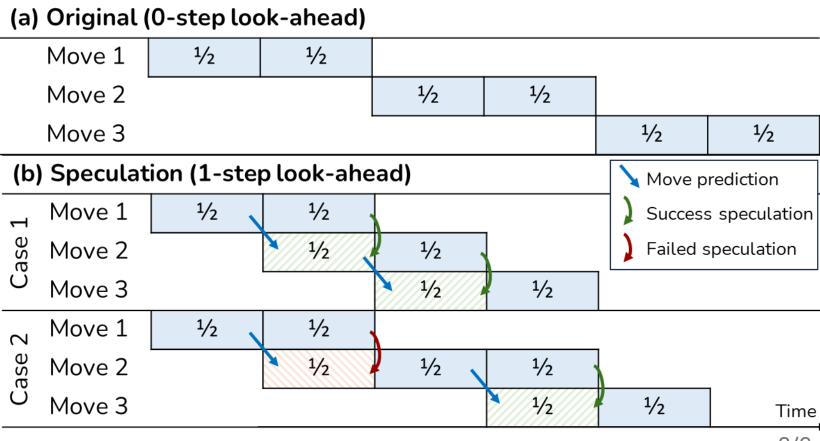
Speculative MCTS Analysis

Based on the speculative pipeline design, we can estimate the latency by solving Markov chain:

$$egin{cases} \mathbf{q}\mathbf{M} = \mathbf{q} \ \|\mathbf{q}\| = 1 \end{cases}$$

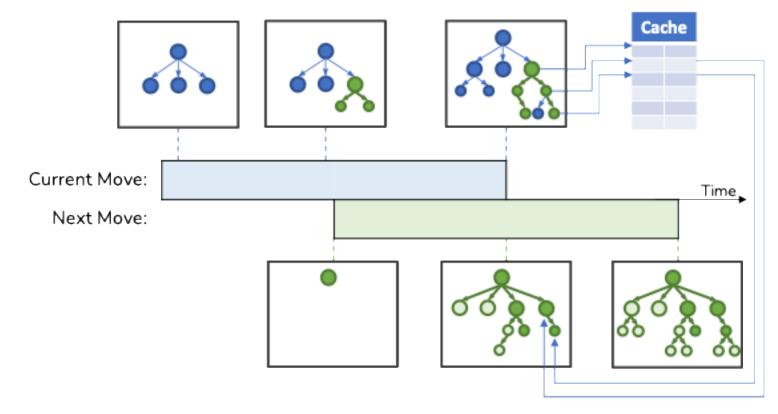
q: steady-state distribution

 ${f M}:$ transition probability



Synergizing Speculation and Caching

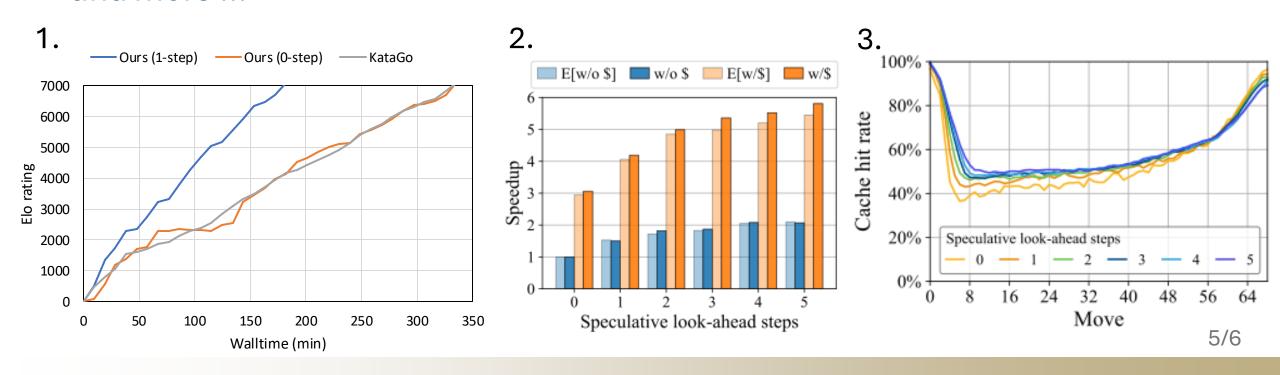
When speculation fails to predict the next move, the inference results remain valuable in the neural network cache.



Source of cache hits: (1) intra-game and (2) inter-game

Evaluations

- End-to-end training: 1.91x speedup compared to the state-of-the-art KataGo program
- 2. Training latency estimation with an average RMSE of 22.2
- 3. Insights into speculation with caching and more ...



For More Details ... Please check our paper!

Why is inter-decision parallelism necessary beyond intergame and intra-decision parallelism?

Why do you choose speculation to achieve inter-decision parallelism?

Can you show a design alternative for Speculative MCTS?

What are the design choices for Speculative MCTS?

What are the limitations of Speculative MCTS?



Thank you

Acknowledgements to:

- National Science and Technology Council, Taiwan
- National Center for High-performance Computing, Taiwan
- Mr. Meng-Yu Tsai

