

# ΩmegaGo

## Elastic, Highly Distributed Go AI Tournaments

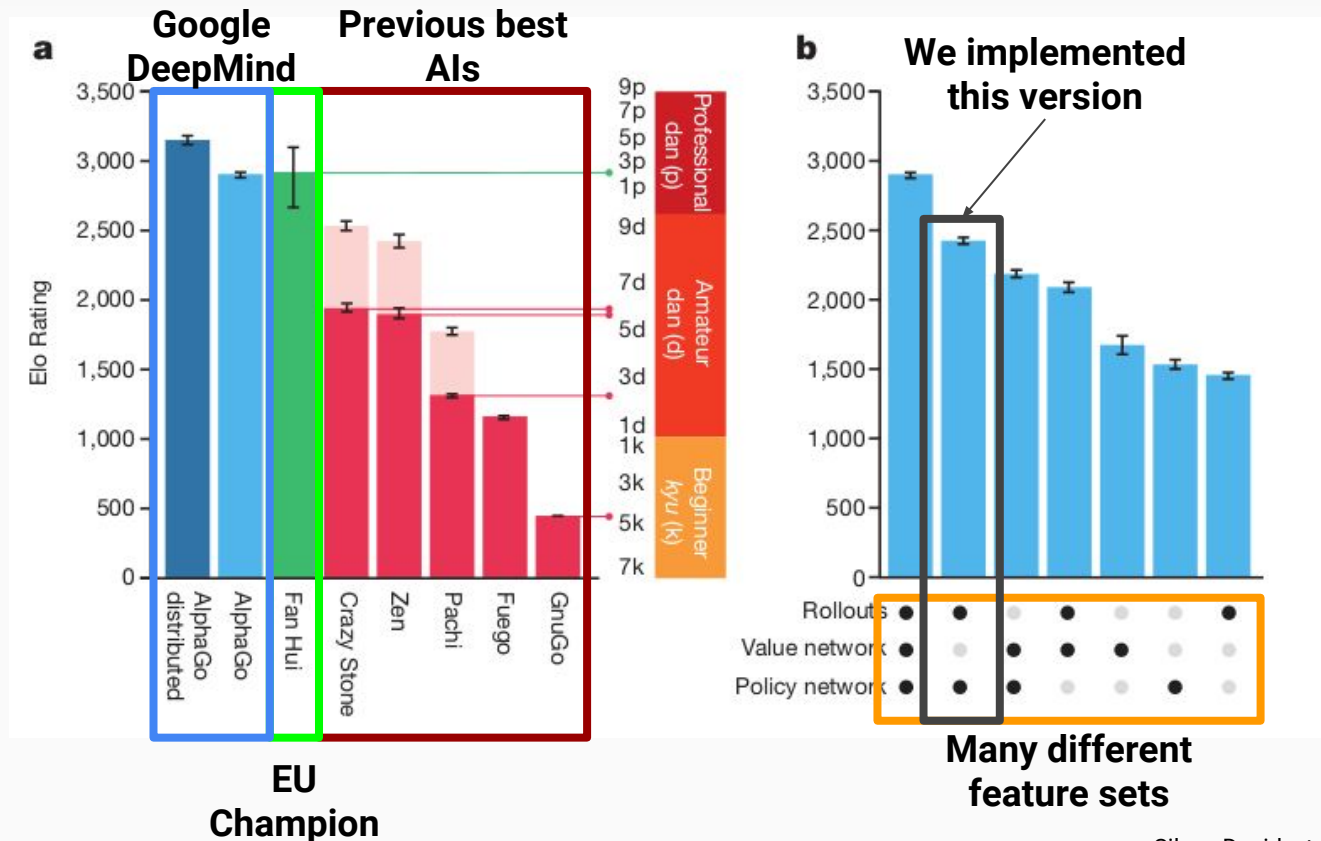
Joe Doyle and Rachel Kositsky  
15-418 S16 Final Project

# Go is like Chess, but harder



- Two players: **black** vs. **white**
- Players place stones on a **19×19** grid
- **Surround** territory and enemy stones to win
- Go:  $\sim 10^{360}$  possible games
- Chess:  $\sim 10^{124}$  possible games

# AlphaGo State of the Art: APV-MCTS

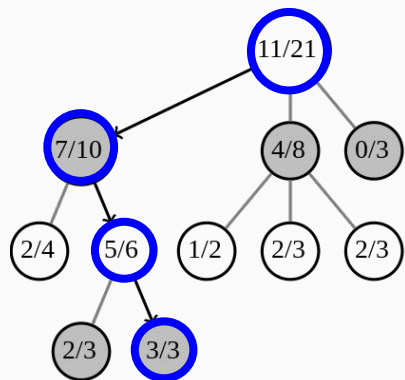


(taller bar = stronger AI)

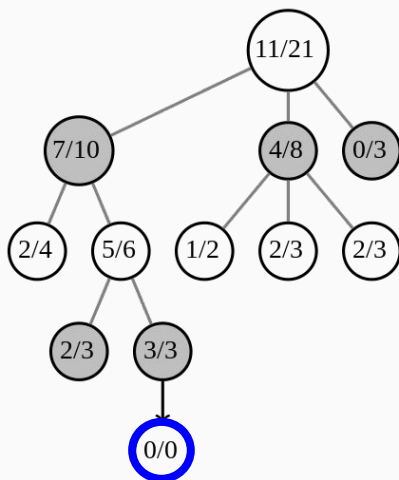
# Lock-Free Monte Carlo Tree Search

- Pick **the best moves you know**
- Play **random moves until the game ends**
- Compute **win/loss**
- **Atomically** update each move's **win probability**

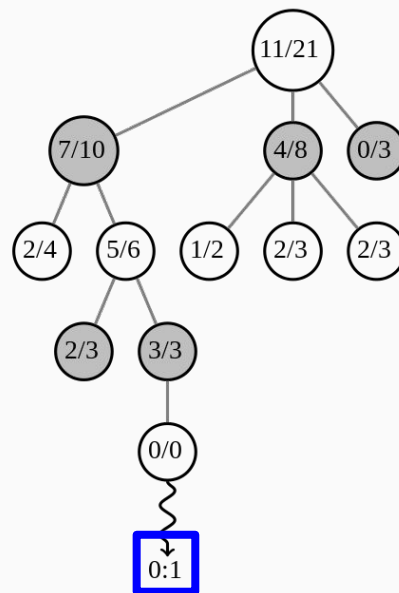
Selection



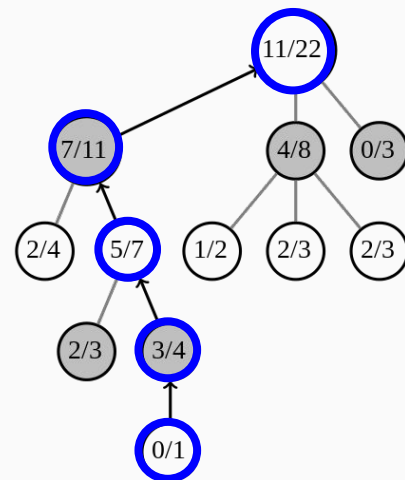
Expansion



Simulation



Backpropagation

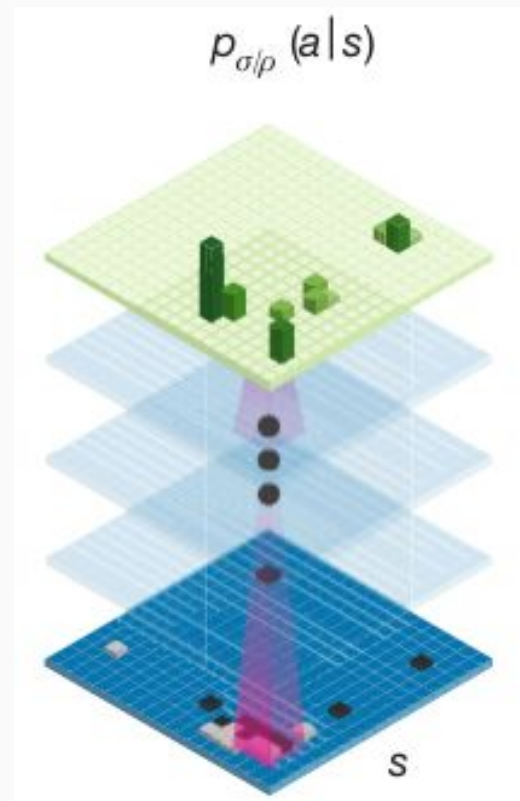


[https://en.wikipedia.org/wiki/Monte\\_Carlo\\_tree\\_search](https://en.wikipedia.org/wiki/Monte_Carlo_tree_search)

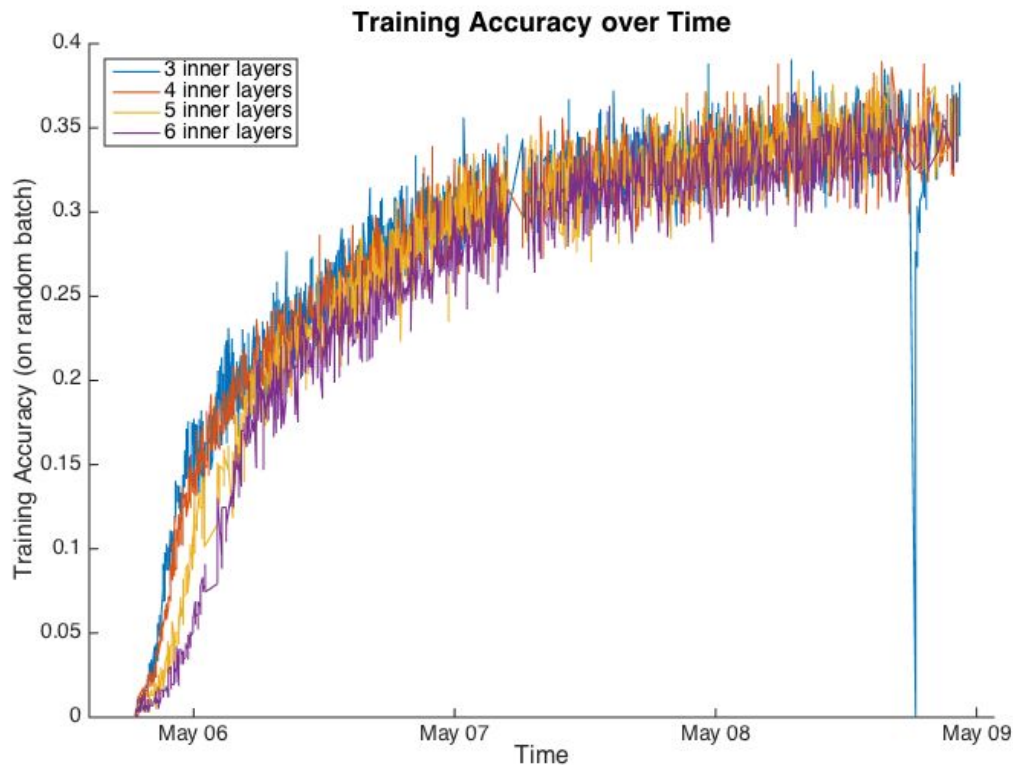
**More threads  $\Rightarrow$  More data  $\Rightarrow$  Higher accuracy**

# Policy Neural Network: Structure

- Inputs: **19×19×46 image**
- Outputs: 19×19 grayscale image
  - **probability of playing** on each point
- $n$  3×3 inner convolutional layers
  - $n = 3, 4, 5, 6$
- **Asynchronous Policy MCTS: Bias MCTS**
  - Policy provides a **hypothesis** for MCTS's experiments



# Policy Neural Network: Training



- Trained on `latelydays`
- 70,000 skilled amateur games
- **Training Accuracy:** % of samples where the network guessed the move played
- AlphaGo: **57%** on this dataset

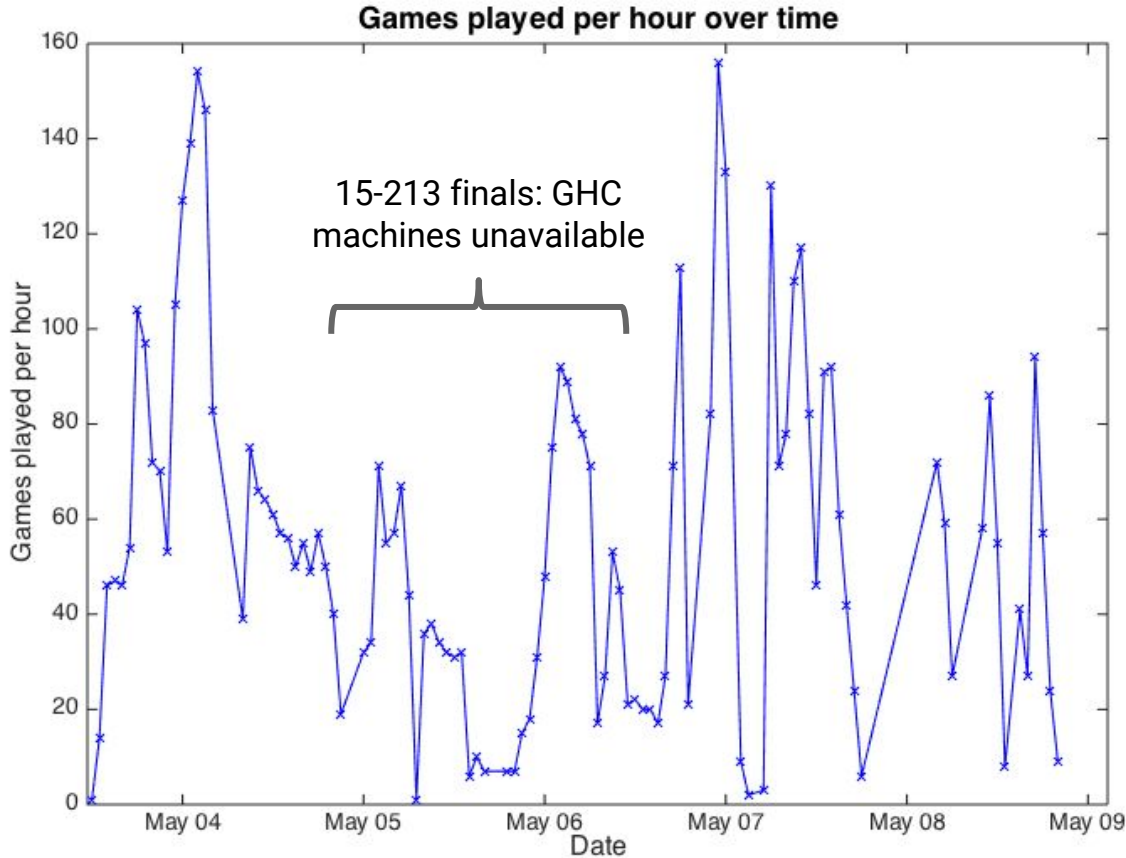
# Comparing AI Skill

# Elastic distributed *tournament* system

- Fully comparing  $n$  AIs requires  $O(n^2)$  games
- **“Use whatever systems you can get your hands on”**
  - `ghc# .ghc.cmu.edu`
- **Scale up and down** with available machines

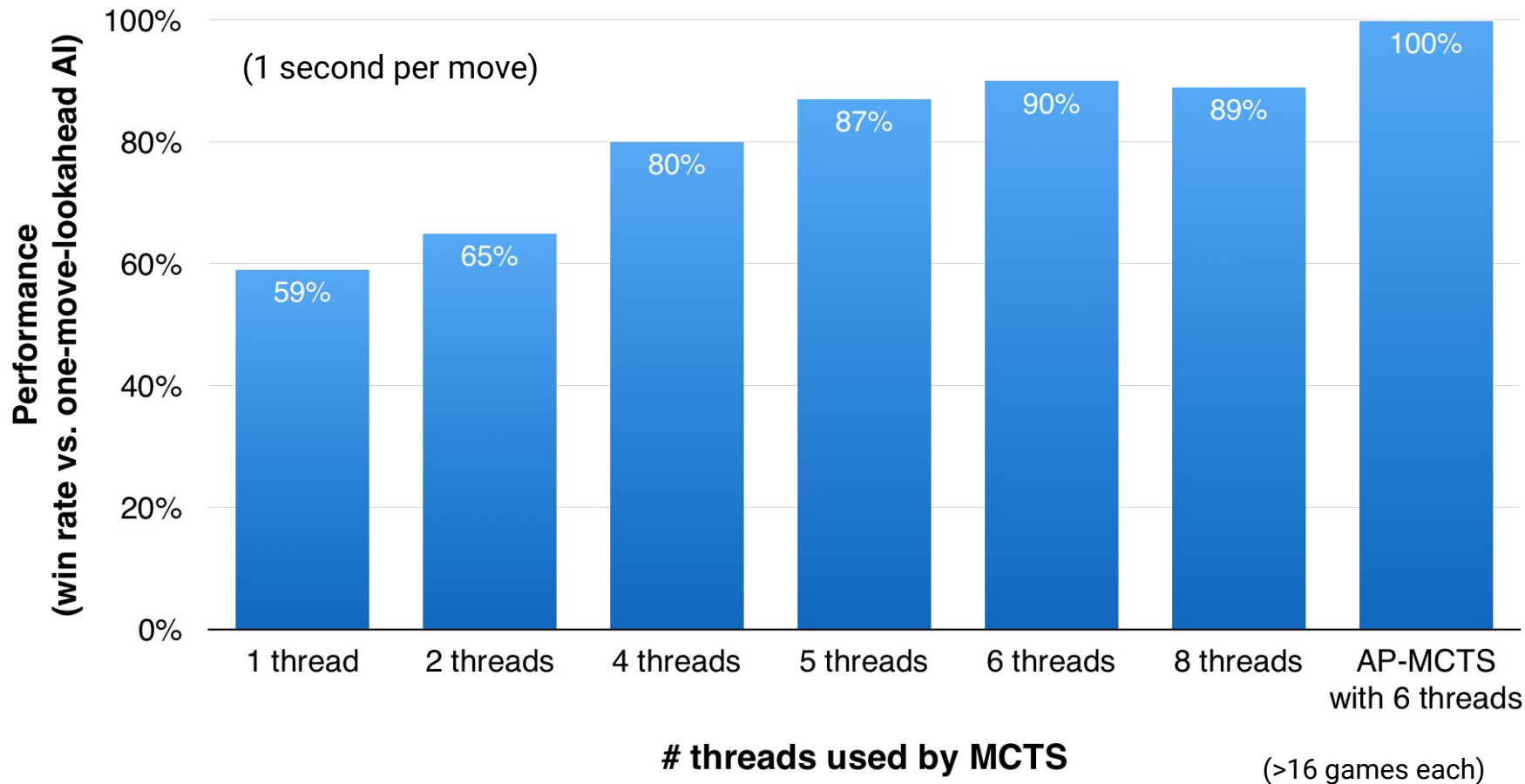


# Elastic distributed tournament system



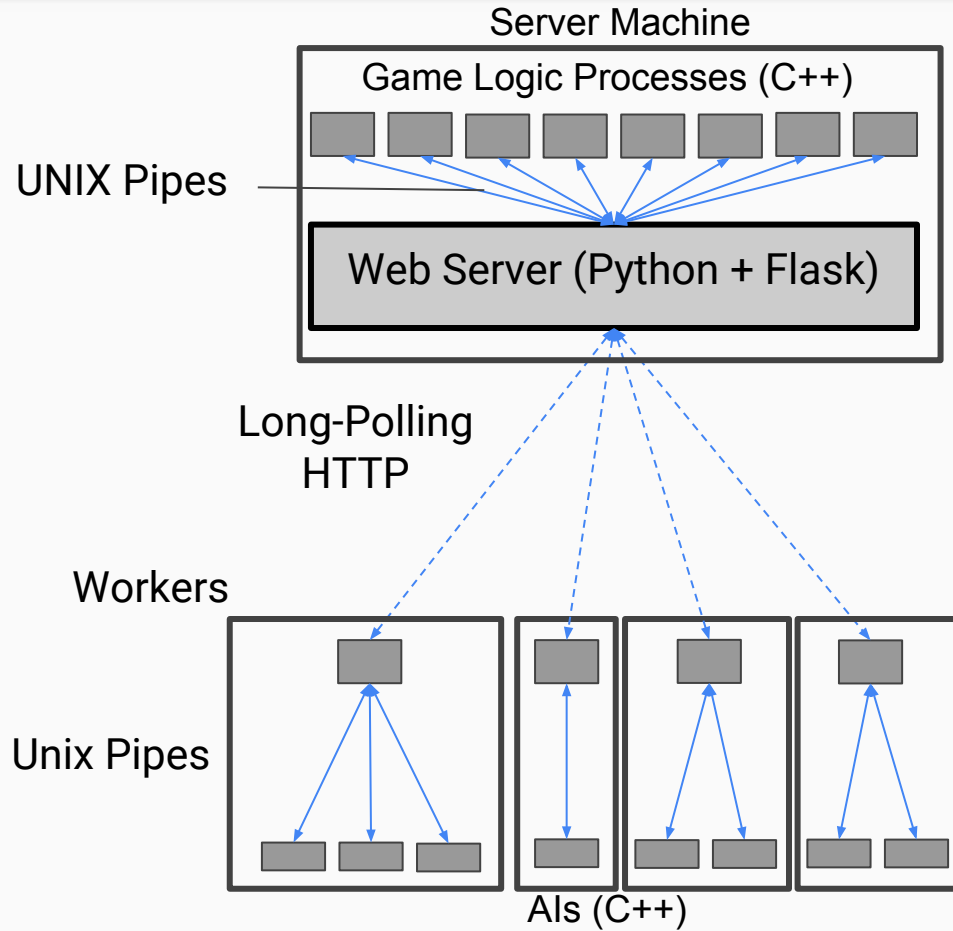
**>5700**  
games played

# MCTS improves with more threads



Demo

# System Architecture



Questions?