Monte-Carlo Search Algorithms

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Outline

- Chess and Go

 Large Game Trees
- Current Monte-Carlo Algorithms
- Current Codebases & Our Additions

 Fuego
 - o Gomba

Chess and Go

Chess

- Long established in AI
- Large Branching Factor,
 - Fixed board size (8x8)
- Games converge to win or loss

Go

- Used in AI More Recently
- Very Large Branching Factor

 Varying Board Size (9x9 to 19x19)
- Games take much longer to converge







Game Trees

- Consist of States and Actions
- Picking an Action leads to a new State
- Find a winning state, choose the action leading to it



Large Game Trees

- Also consist of States and Actions
- How do you find the best action?
- How do you find one close to best?

Winning

State

Losing

States

UCT

- Upper Confidence Trees
- "Multi-Armed Bandit"
 Simulated Regret
- Property of "Upper Confidence Bounds"
 Balances Exploration Through Tree
- What this means
 - $\circ\,$ Always exploring the current best move
 - Converges towards optimal choice

UCT-RAVE

- Rapid Action Value Estimation (RAVE)
- Extension of basic UCT
- Updates the values across multiple states, rather than maintains the value on a per-action-state basis.
- AMAF (all moves as first) is a general name for this type of heuristic.

UCT-RAVE



Implementation - Fuego

- Pre-Existing Library of Go Algorithms
- UCT-RAVE already implemented
- Uses Go Text Protocol (GTP)to play against
 Other Go Programs
 GnuGo, MoGo
 - o Humans
 - GoGui
- First program to defeat a 7 Dan Go player in 9x9 Go.

Fuego - Framework

- Large Codebase
- Only One External Library
- Multi-Threaded
- Open Source
- Extendable



Fuego - Experiments



Time = 2T

Time = T + t

Disadvantages

Fuego et al. take a very long time to test. Play against separate algorithm for many games Change parameters Play again

Effectiveness of a new algorithm difficult to prove. Parameters difficult to optimize.

Cluster Jobs / Parallel Testing improves response of results

There is an alternative.

Artificial Game Tree

- Faster speed
- Better heuristic
- Easily twisted parameters (branching factors, depth, etc.)

Gomba

- Developed by Daniel Bjorge and John Schaeffer in 2010
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- Mini-max Artificial Game Tree
- Lazy State Expansion
- Fast Action Simulation
- Fast Termination Evaluation
- Fast Heuristic Evaluation

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Problem with Gomba

- AMAF algorithms UCT RAVE algorithm
 - transposition table
 - $_{\odot}\,$ global knowledge of the moves
- Structure of Gomba cannot support this type of algorithms

Our improvements

- Modified the tree structure to better support AMAF algorithms
- Add global knowledge of the moves
- Better correlation and consistency among the moves
- Modification to the lazy state expansion

Correlations among the moves



Adjust to the tree structure



Original lazy state expansion

New lazy state expansion

Improvements



Correlations



Conclusions & Future Work

- Improved Gomba testing framework

 Better support of AMAF algorithms (UCT RAVE)
- Implemented RAVE-Pool in Fuego

 With Variations

Köszönöm!

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