Monte Carlo Search Tree and Its Applications

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Kasparov vs Deep Blue





Kasparov vs Deep Blue

Great display of artifical intelligence (AI) Techniques employed by IBM

- Brute force deterministic approach
- Human knowledge

Limitation

Scalability into larger search spaces

Monte Carlo tree search (MCTS) is an alternative method



Outline

Introduction

Naive MCTS Implementation

Applying MCTS to Go

Applying MCTS to Narrative Generation

Conclusion



Monte Carlo Tree Search (MCTS)

- Combines random sampling and game trees
- Lightweight random simulations
- Probabilistic not deterministic
- Useful for problems with larger search spaces



Applying MCTS to Go

Go

- Board game about positional advantage
- Game board for Chess:
 - ▶ 8x8
- Average possible configurations for a game of Chess:
 - ▶ 10¹²⁰
- Game board for Go:
 - ▶ 19x19
- Average possible configurations for a game of Go:
 - ► 10⁷⁶¹



Applying MCTS to Narrative Generation

- Useful Applications
 - Video game replay value
 - Educational applications
- The search space scales with the number of characters, items, locations, and actions



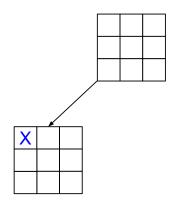
Naive MCTS Implementation

Applying MCTS to Go

Applying MCTS to Narrative Generation

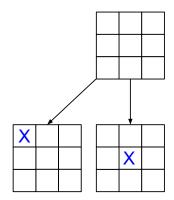


TicTacToe Diagram

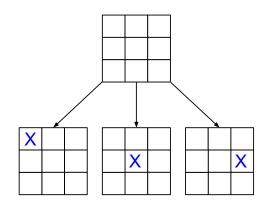




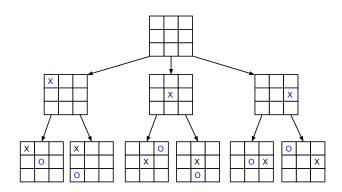
TicTacToe Diagram





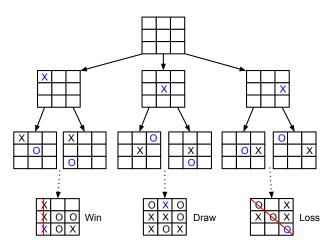






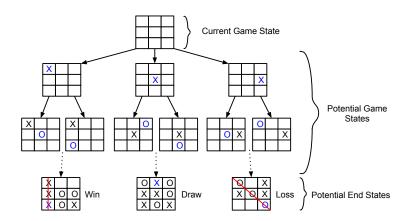


TicTacToe Diagram



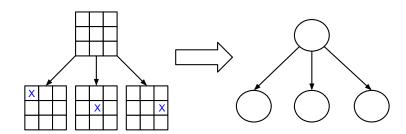


TicTacToe Diagram

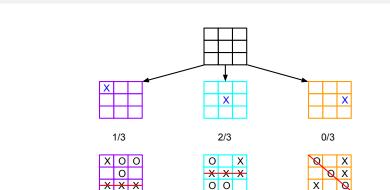




Tree Structure

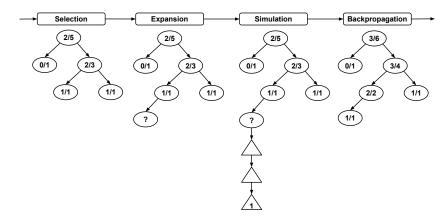






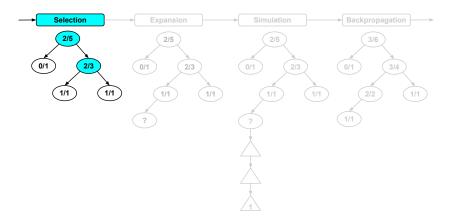
X O X



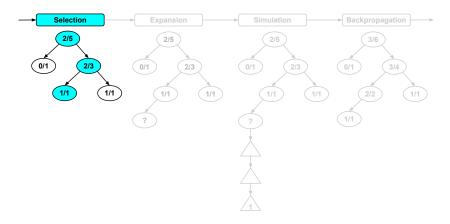




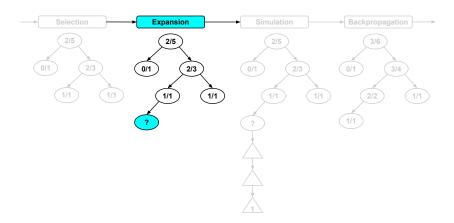




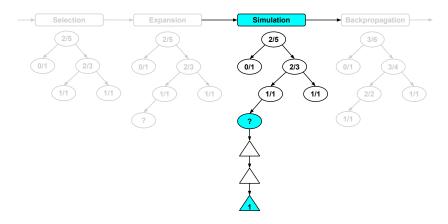




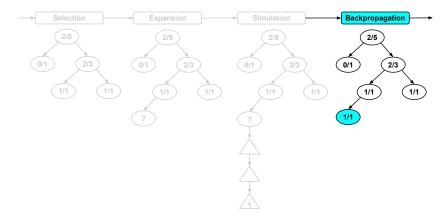




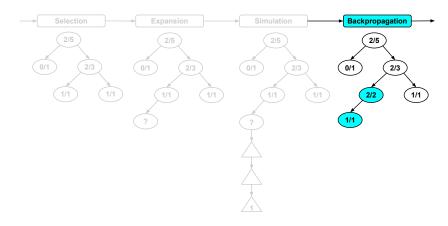




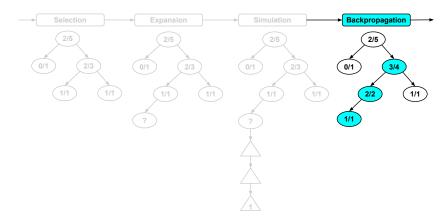




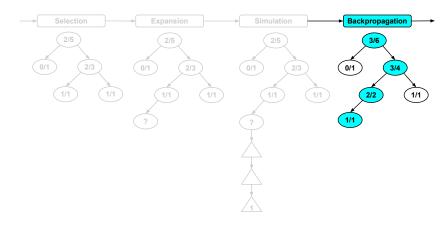




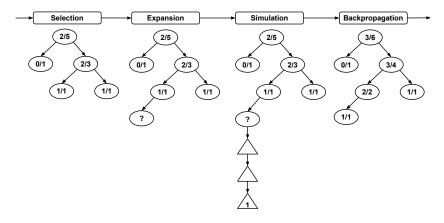




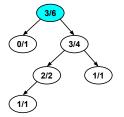


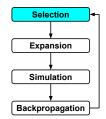


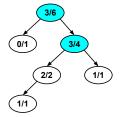


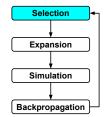


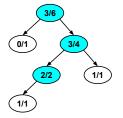


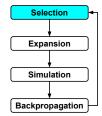










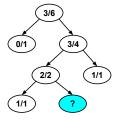


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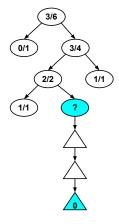
Expansion

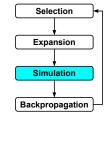
Simulation

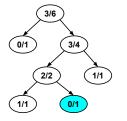
Backpropagation

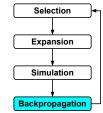










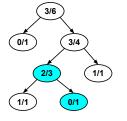


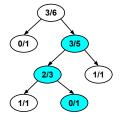
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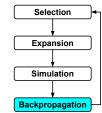
Expansion

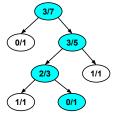
Simulation

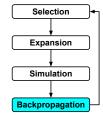
Backpropagation

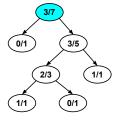


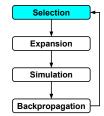


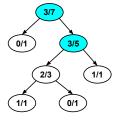


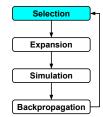


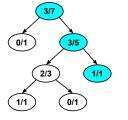


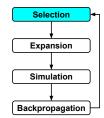


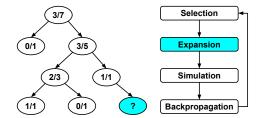




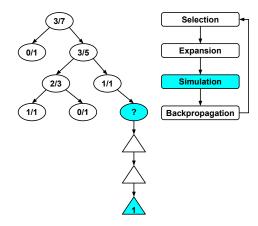




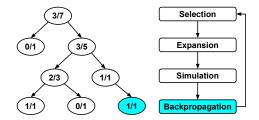




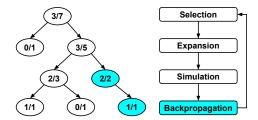




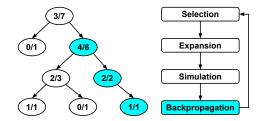




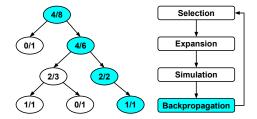
Four Steps Diagram











What Happens When We Choose a Move?

Now we have:

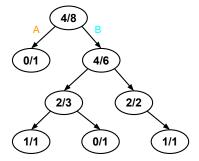
A tree structure

Naive MCTS Implementation

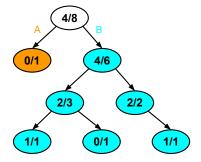
A method of generating the tree

What happens when we need to choose a move?



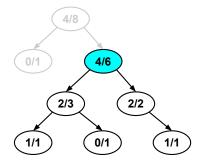






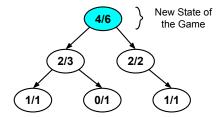


Choosing a Move





Choosing a Move





- We might overlook better paths
- Exploration vs Exploitation
 - Exploration looks at more options
 - Exploitation focuses on the most promising path
- Must find a balance between the two



Upper Confidence Bound Applied to Trees (UCT)

$$UCT(node) = \underbrace{\frac{W(node)}{N(node)}}_{\text{Value of the Node}} + \underbrace{\sqrt[C]{\frac{In(N(parentNode))}{N(node)}}}_{\text{Exploration Bonus}}$$

- W represents the number of simulated wins
- N represents the total number of simulations
- C is an experimental constant
- Used during tree traversal
- Balances exploration vs exploitation



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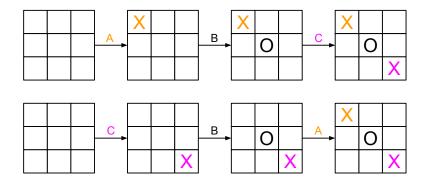
MCTS applied to Go

What variations can we make specific to Go? In Go each player takes turn placing pieces on a game board

- How much does the order of these moves matter?
- Can we use this to improve MCTS in the context of Go?



Tree Redundancy



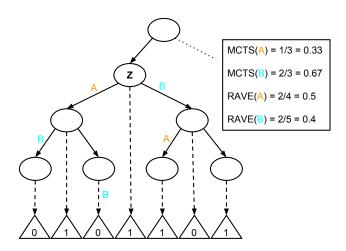


Rapid Action Value Estimate (RAVE)

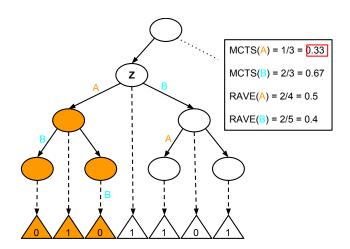
- Takes advantage of tree redundancy
- Moves have no contextual dependencies
- Stores the value of a move within a subtree at each node



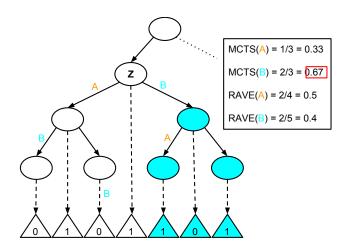
RAVE Diagram





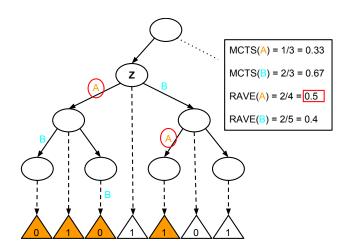




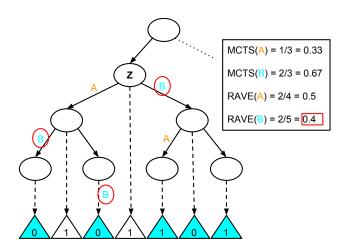




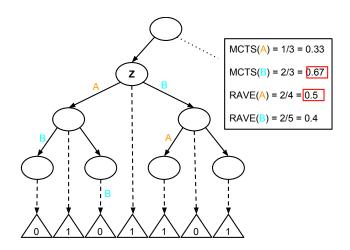
RAVE Values













- Very powerful approach
- Each simulation provides us with more information
- Sometimes we do need contextual dependencies
 - Example: Close tactical battles



Combines MCTS values with RAVE values

- Uses a weighted average
- Favors RAVE values when fewer simulations have been performed
 - Contextual dependencies are unknown
- Favors MCTS values when more simulations have been performed
 - Contextual dependencies are more developed



- Deterministic approaches could hardly defeat low level amateurs
- Computer Go programs use MC RAVE
 - MoGo
 - Crazy Stone
- Can compete against top pros in 9x9 Go
- Can compete against top pros in handicapped 19x19 Go



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Narrative Generation

Kartal et al. applied MCTS to Narrative Generation

- Crime story
- User defines the set up and goals for the story
 - Example Setup: The detective starts in his office
 - Example Goal: The killer must be arrested

Unlike Go and other games

- Slightly different tree structure
- Evaluation function needed



Actions

- Actions drive the story
- Actions are believable based on context
 - Example: Inspector searches for clues
 - Example: Character A kills Character B



Example Actions

Move(A, P): A moves to place P.

Kill(A, B): B's health to zero(dead).

Earthquake(P): An earthquake strikes at place P.

- Actions take the place of moves as nodes
- No clear end state
- The researchers used a set threshold during simulation



Evaluation function

- Method of giving nodes value
- Incorporates believability and goal completion
- Ensures stories are interesting

Value(story) = Believability(story) * GoalCompletion(story)

- Believability is the mathematical product of every action in a story
- The value is between 0 and 1



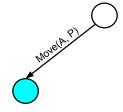
Narrative Generation Test

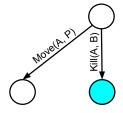
MCTS compared against three deterministic algorithms

- Breadth-first search
- Depth-first search
- Best-first search

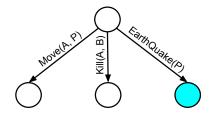


Breadth-First Search

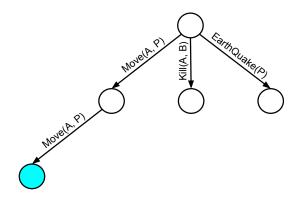




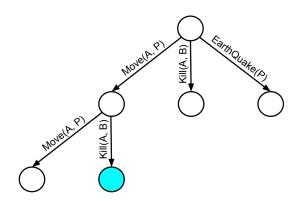




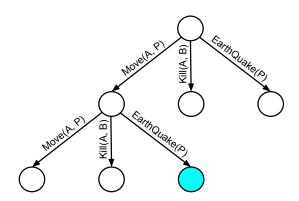






















Depth-First Search









Depth-First Search



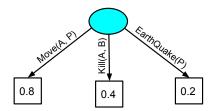


Depth-First Search

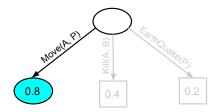




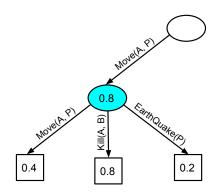
Best-First Search





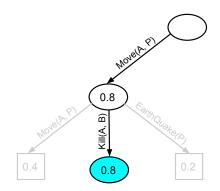






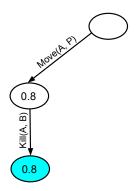


Best-First Search





Best-First Search





Test Conditions

Goals for the narrative:

- At least two people are killed
- The killer is arrested

Each algorithm was given two budgets

- ▶ 100,000 nodes
- 3 million nodes

Each algorithm ran three times The score of the narratives were averaged



	MCTS	Breadth- first	Depth- first	Best- first
Low Budget	0.07	0.05	<0.001	0.005
High Budget	0.9	0.06	<0.01	<0.01

- MCTS performed the best in both
- Breadth-first came the closest out of the deterministic algorithms



Stories Produced by MCTS

- Stories from MCTS tended to be believable
- Completed both user defined goals
- Some Problems
- Overall reasonable narratives



Low Scoring Example from Breadth-First

Sherlock moved to Alice's House. An Earthquake occurred at Alice's House! Sherlock and Alice both died due to the earthquake.



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Conclusion

- MCTS successful in extending AI capabilities
- Tackles problems with larger search spaces
- Effective in Go and Narrative Generation.
- Applicable to other problems
 - Can outperform humans in many puzzles
 - Real time games
 - Super Mario Brothers



Any Questions?





Conclusion