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# MLDS 422 - Intro to Python

## Lab 7

Sungsoo Lim  
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NORTHWESTERN  
UNIVERSITY

# Today's Lab Materials

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- ▶ Monte Carlo Tree Search
  - ▶ Algorithm Background
  
- ▶ Implementation Details

# Monte Carlo Tree Search



Figure: AlphaZero

- ▶ <https://www.analyticsvidhya.com/blog/2019/01/monte-carlo-tree-search-introduction-algorithm-deepmin>

# Monte Carlo Tree Search

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- ▶ With Go, the number of possible moves given a board position is very large
- ▶ AlphaZero simulates ahead from a given board position to the end of game (sample all possible next moves of all future board positions)
- ▶ The algorithm selectively selects from the possible next moves - 'promising ones' that lead to victory
- ▶ Based on the simulations, it calculates the win probabilities from the given position for each possible move and takes the highest move based on win probability

# Monte Carlo Tree Search

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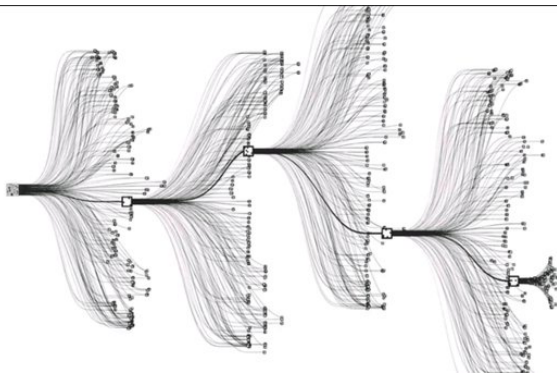


Figure: [https://www.quora.com/](https://www.quora.com/What-does-it-mean-that-AlphaGo-relied-on-Monte-Carlo-tree-search)

What-does-it-mean-that-AlphaGo-relied-on-Monte-Carlo-tree-search

# Monte Carlo Tree Search

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- ▶ Selection - from the root node, select a leaf node (this step needs to be selective - more promising moves are selected more often)
- ▶ Expansion - from the selected leaf node, add all its child nodes to the tree
- ▶ Simulation - from one of the child nodes, simulate to end of game
- ▶ Backpropagation - for the path from the root node to the child node, update their statistics
- ▶ Statistics - for the case of Go - did the game end in a win, tie, or loss (1, 0, -1). How many times did the path get visited?

# Monte Carlo Tree Search

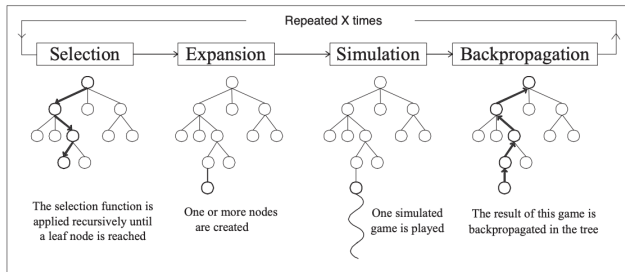


Figure 1: Outline of a Monte-Carlo Tree Search.

Figure: Sutton and Barto 2009

# Monte Carlo Tree Search - Selection

- ▶ Upper Confidence Bounds algorithm (Auer *et al.* 2002)
- ▶  $a$  - action (which move should be done to select the leaf node from the root node)
- ▶  $A_s$  - the set of all possible moves from the root node  $s$
- ▶  $Q(s, a)$  - the statistics we store (win, tie or loss) - if we select  $a$  from  $s$ , did it lead to win, tie or loss?
- ▶  $c$  - a hyperparameter to balance the first term and the second term
- ▶  $n_s$  - how many times the give root node was visited
- ▶  $n_{s,a}$  - how many times the move  $a$  was selected from  $s$

$$a^* = \arg \max_{a \in A_s} \left[ Q(s, a) + c \sqrt{\ln(n_s) / n_{s,a}} \right],$$