

General Game Playing

Statistical Search

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Statistical Methods

Evaluation functions examined thus far

Based on properties of small portions of game tree
e.g. mobility, intermediate state values

Statistical Methods

Based on statistical analysis

Based on larger portions of game trees

Methods:

Monte Carlo Search

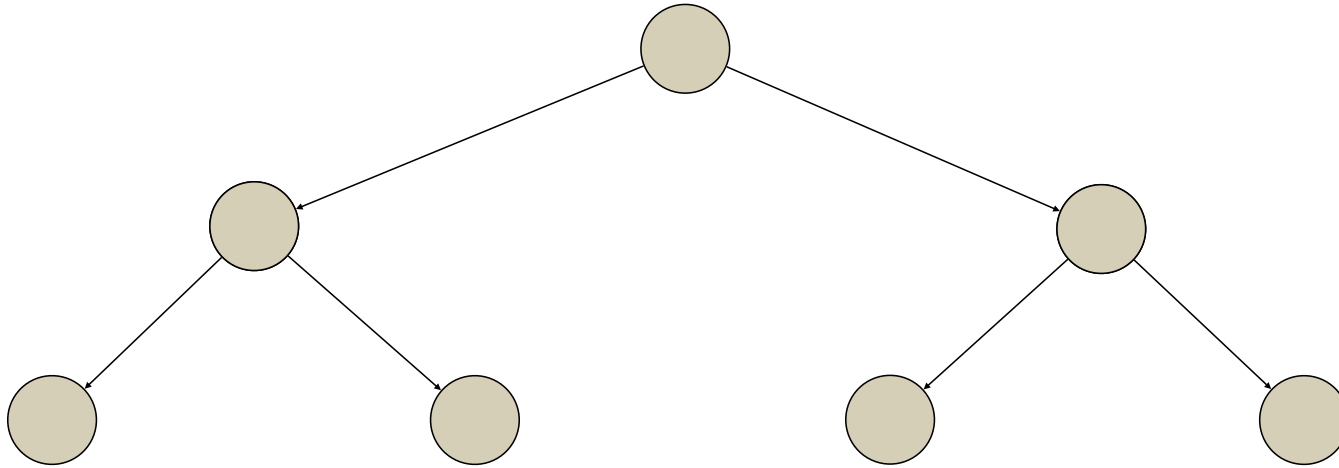
Monte Carlo Tree Search (in particular UCT)

Monte Carlo Search

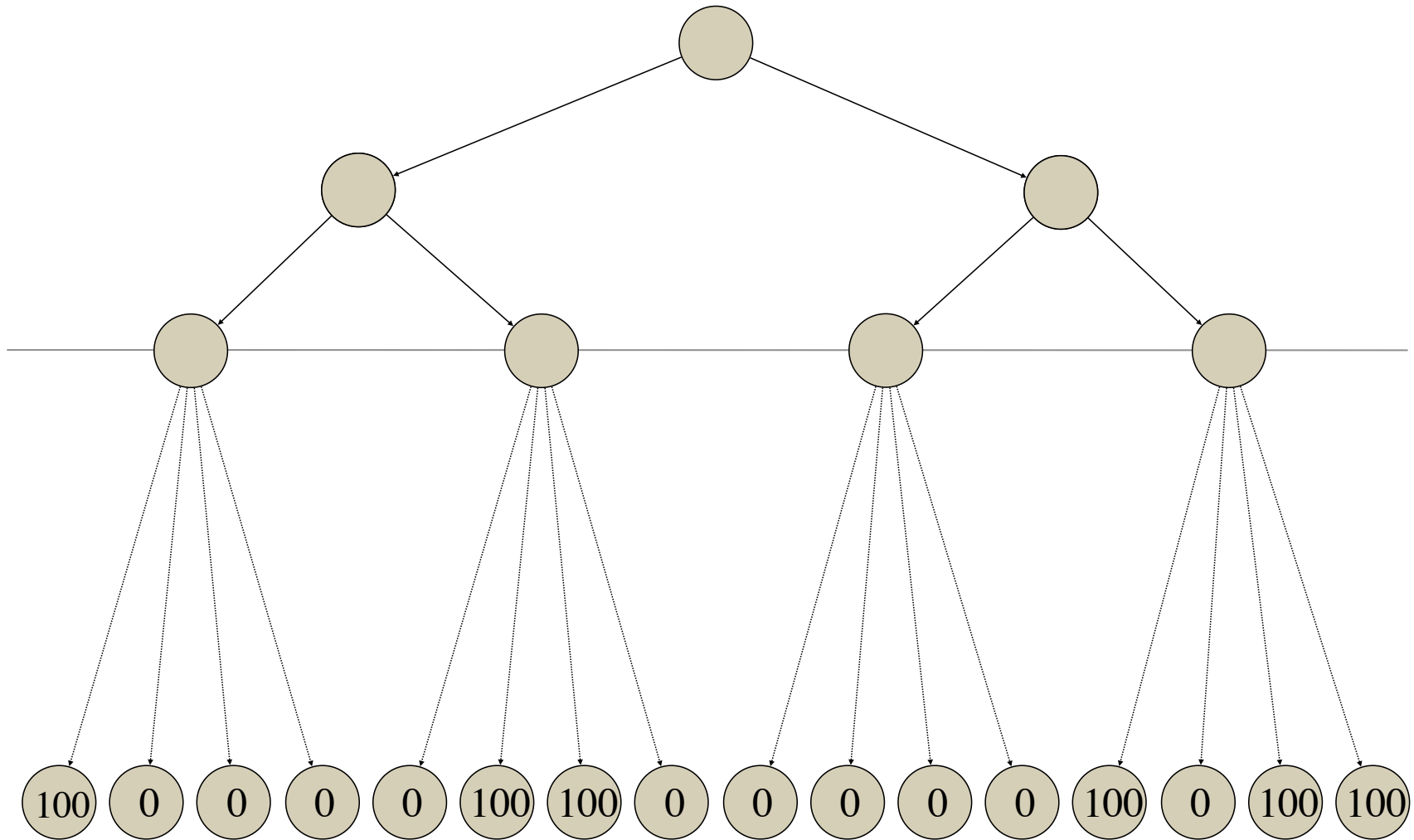
Basic Idea

- (1) Optionally explore game graph to some level.
- (2) Beyond this, explore to end of game from fringe nodes, making random choices for moves of all players.
- (3) Assign expected utilities to fringe states by summing utilities and dividing by number of trials.

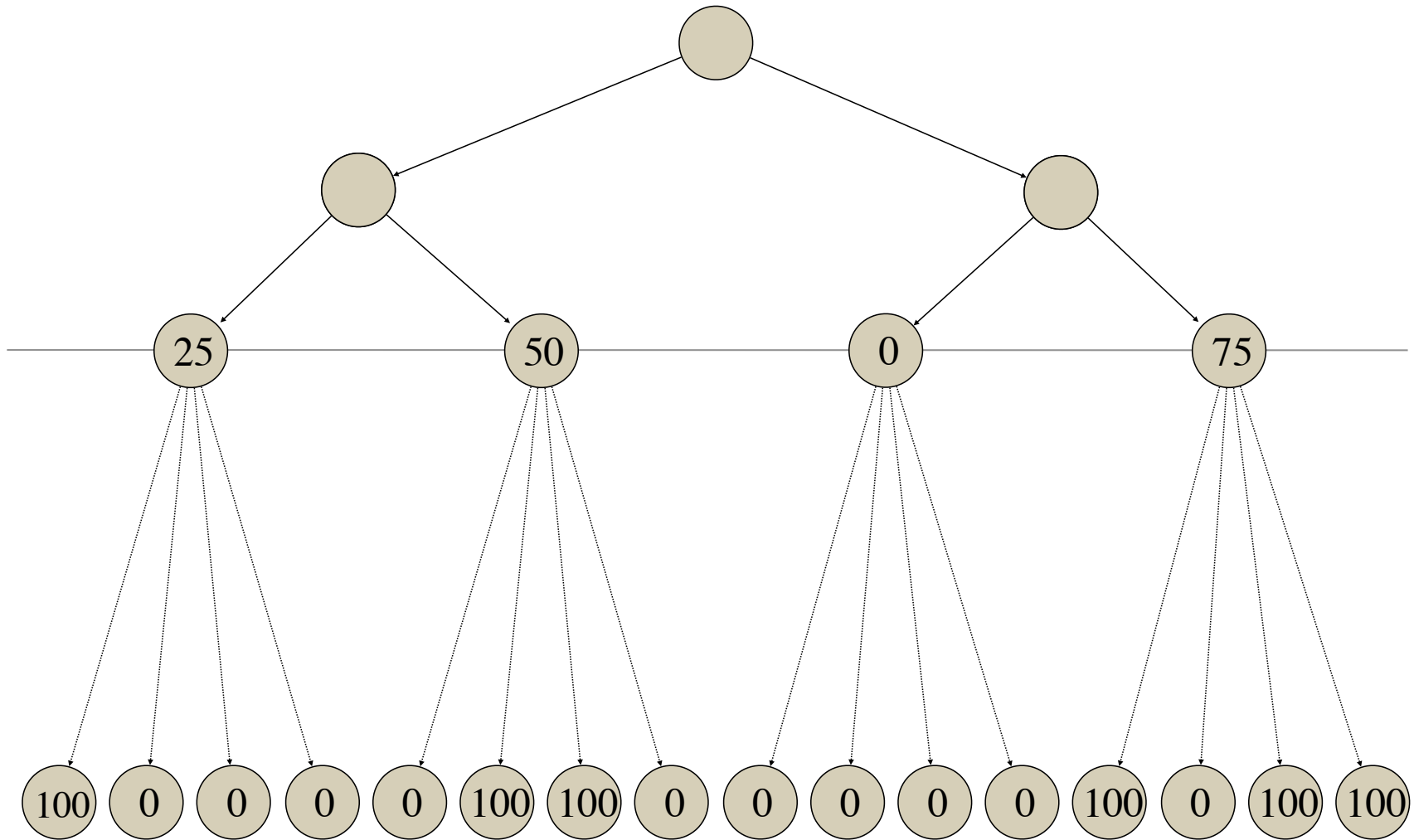
Example



Example



Example



Implementation

```
function maxscore (state,level)
{if (findterminalp(state,ruleset))
    {return findreward(role,state,ruleset)};
  if (level>levels) {return montecarlo(state)};
  var actions = findlegals(role,state,ruleset);
  var score = 0;
  for (var i=0; i<actions.length; i++)
    {var result = minscore(actions[i],state,level);
     if (result==100) {return 100};
     if (result>score) {score = result}};
  return score}
```


Implementation

```
function montecarlo (state)
  {var total = 0;
   for (var i=0; i<count; i++)
     {total = total + depthcharge(state)};
   return total/count}

function depthcharge (state)
  {if (findterminalp(state,ruleset))
    {return findreward(role,state,ruleset)};
   var move = seq();
   for (var i=0; i<roles.length; i++)
     {var options = findlegals(roles[i],state,library);
      var best = randomindex(options.length);
      move[i] = options[best]};
   var newstate = findnexts(move,state,library);
   return depthcharge(newstate)}
```

Problems and Features

Problems

Optimistic - opponent might not respect probabilities

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Does not utilize game structure in any useful way

Problems and Features

Problems

- Optimistic - opponent might not respect probabilities
- Does not utilize game structure in any useful way

Benefits

- Fast because no branching
- Small space because nothing stored in probes
- Provides guidance when other heuristics fail

Monte Carlo Tree Search

Basic Idea

Monte Carlo Tree Search (MCTS) is a search method that relies on random probes to estimate state values. Blend of Greedy and MCS.

Like MCS:

- Builds up game tree incrementally

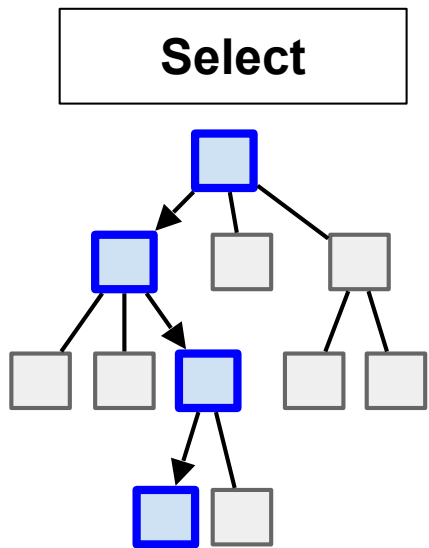
- Random probes to end of game to estimate state values

Like Greedy:

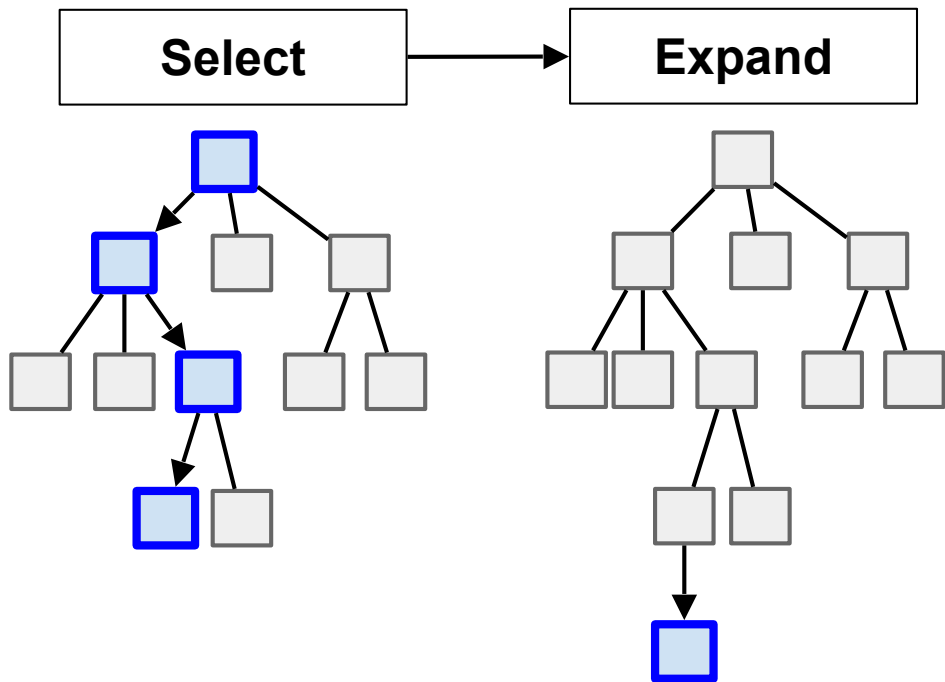
- MCS expands tree uniformly

- MCTS expands nonuniformly like Greedy

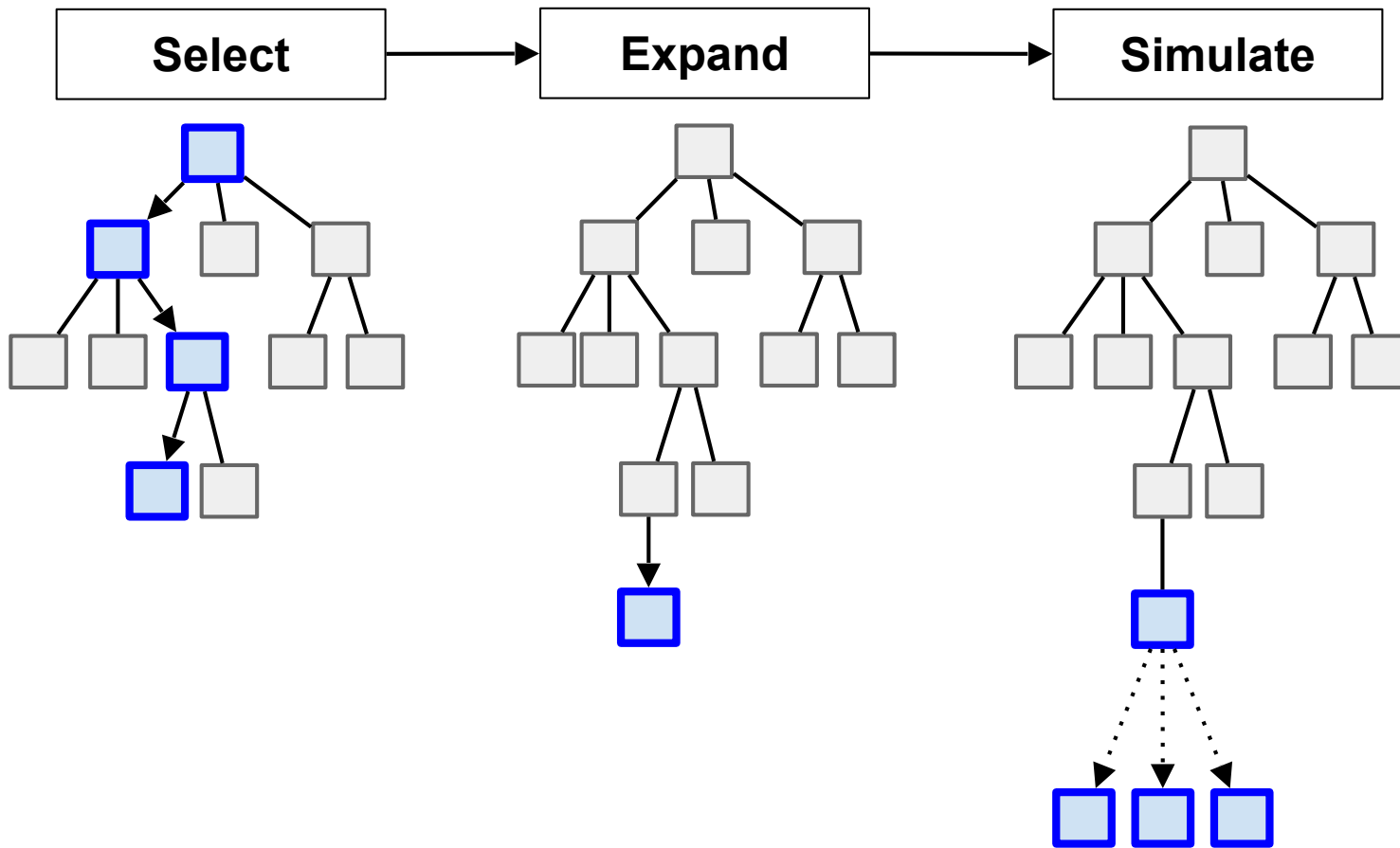
Overview



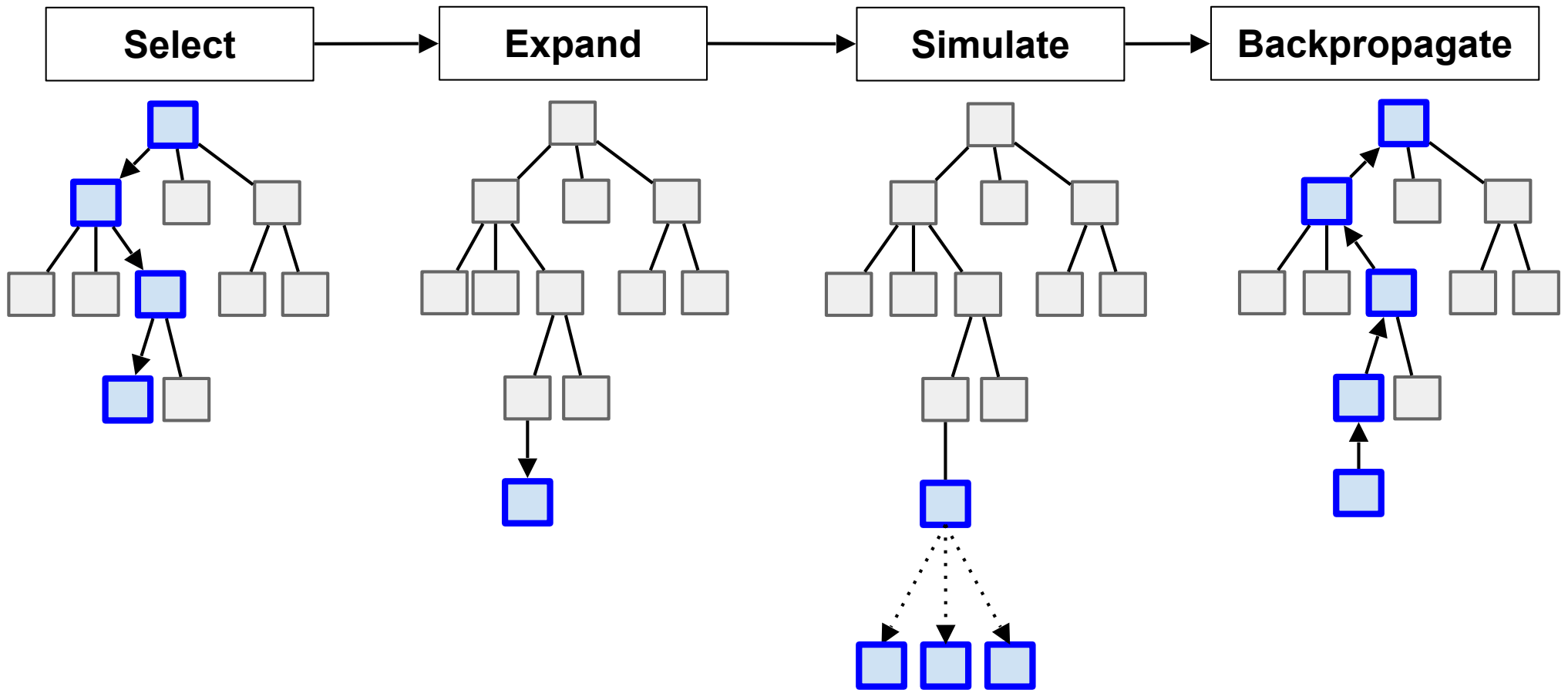
Overview



Overview

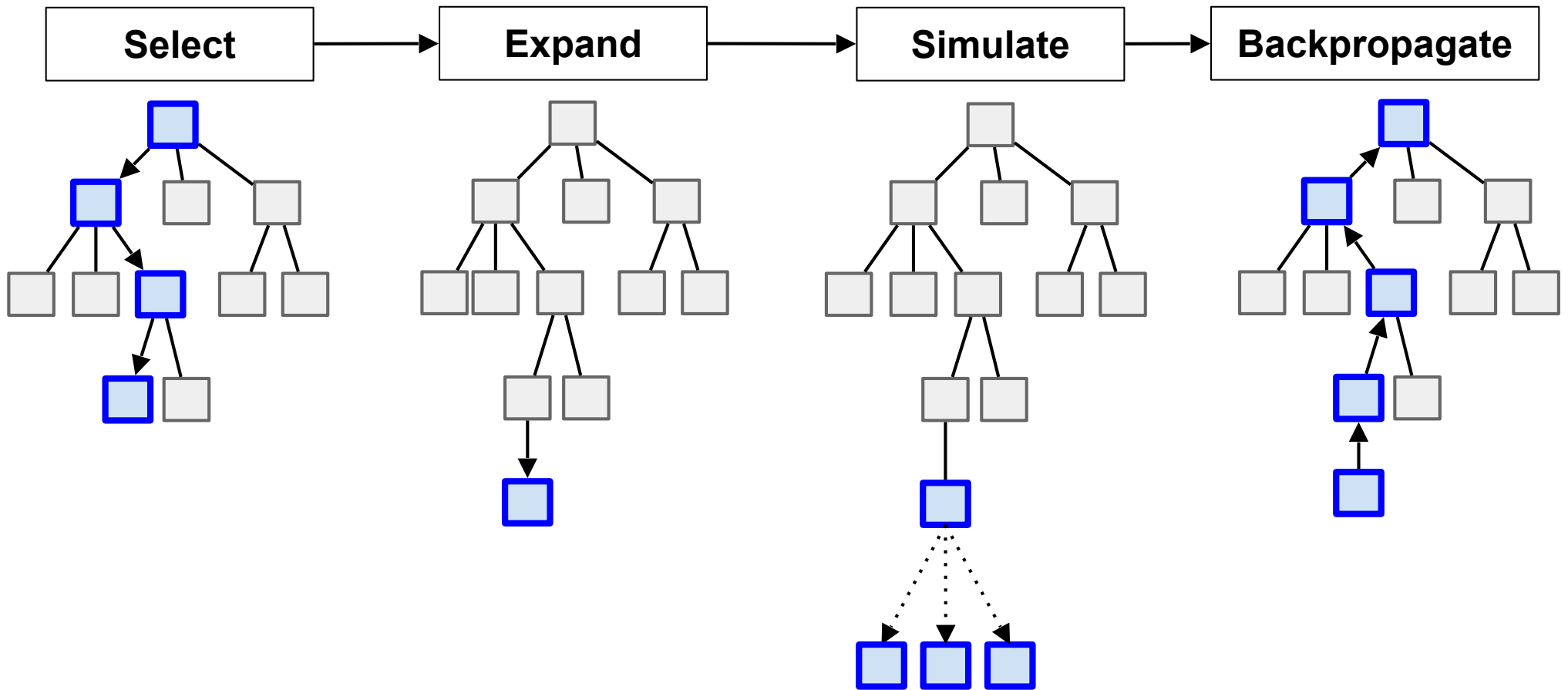


Overview

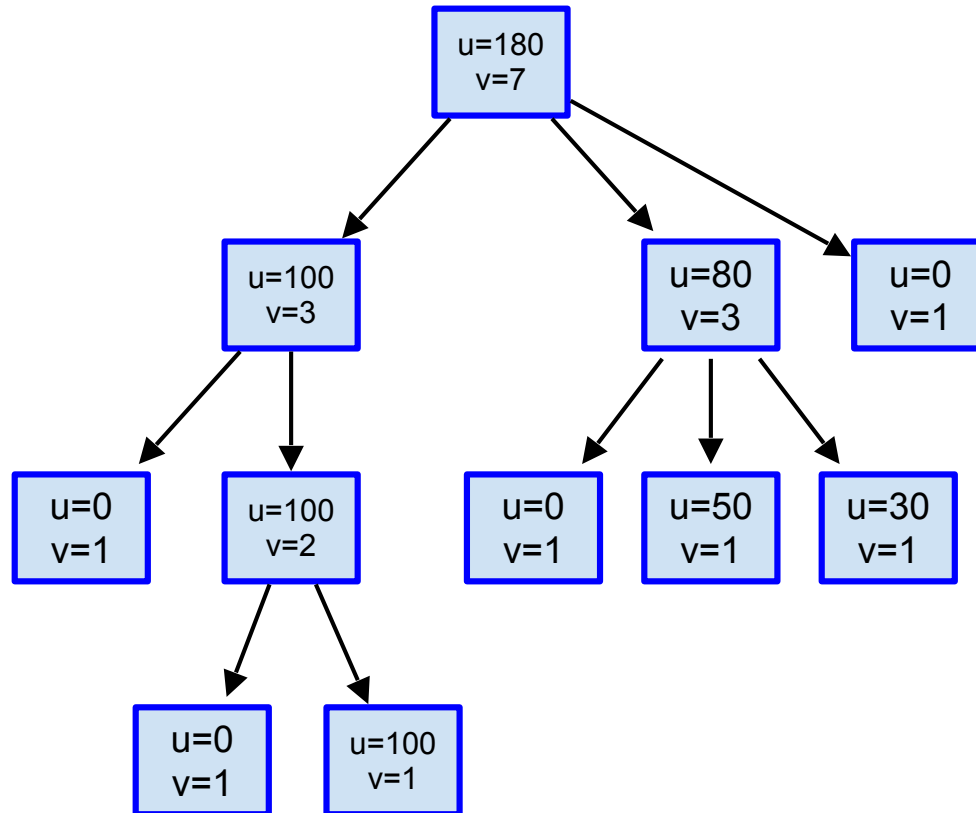


Overview

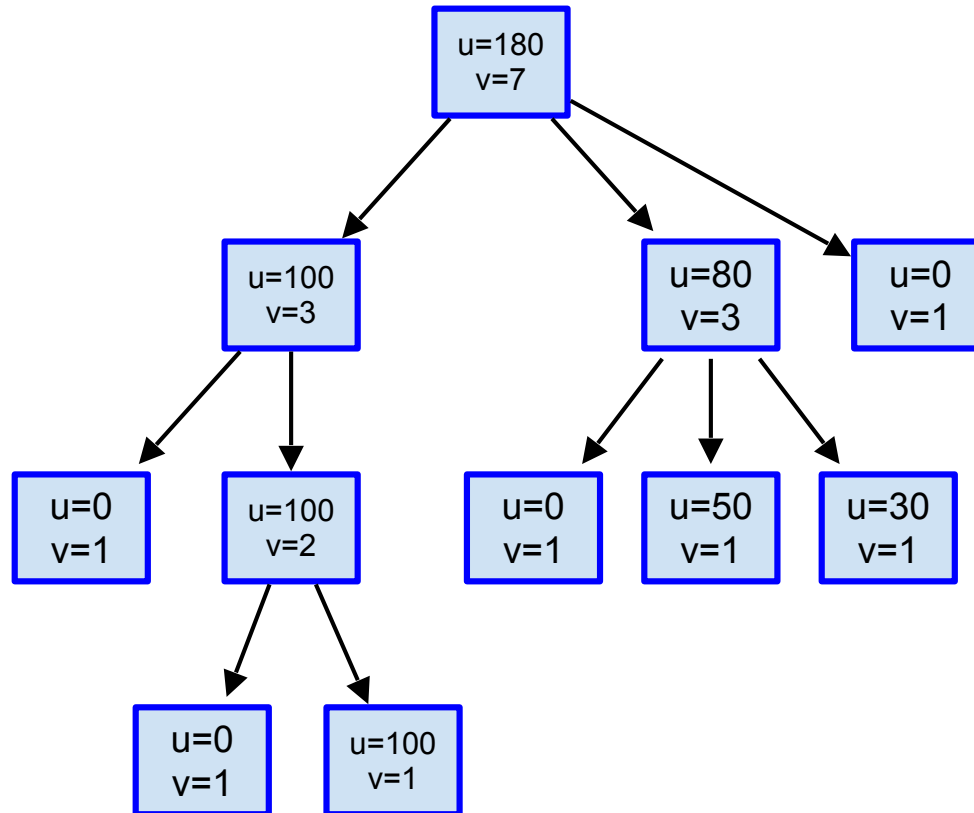
Repeat until timeout



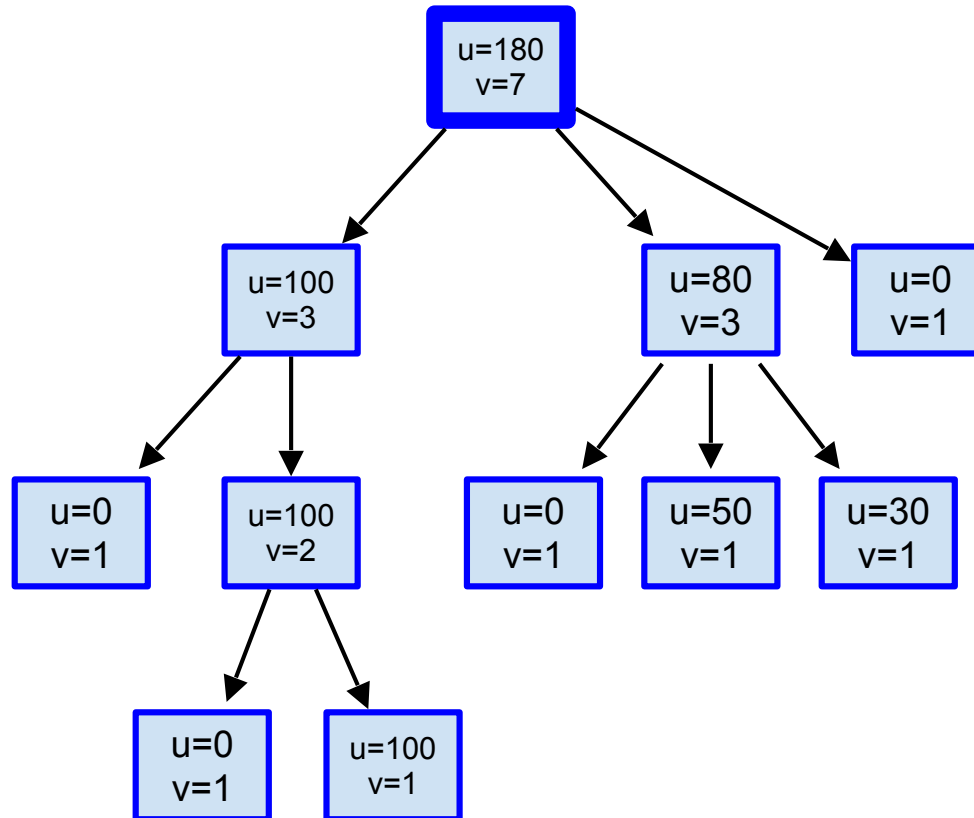
Single Player



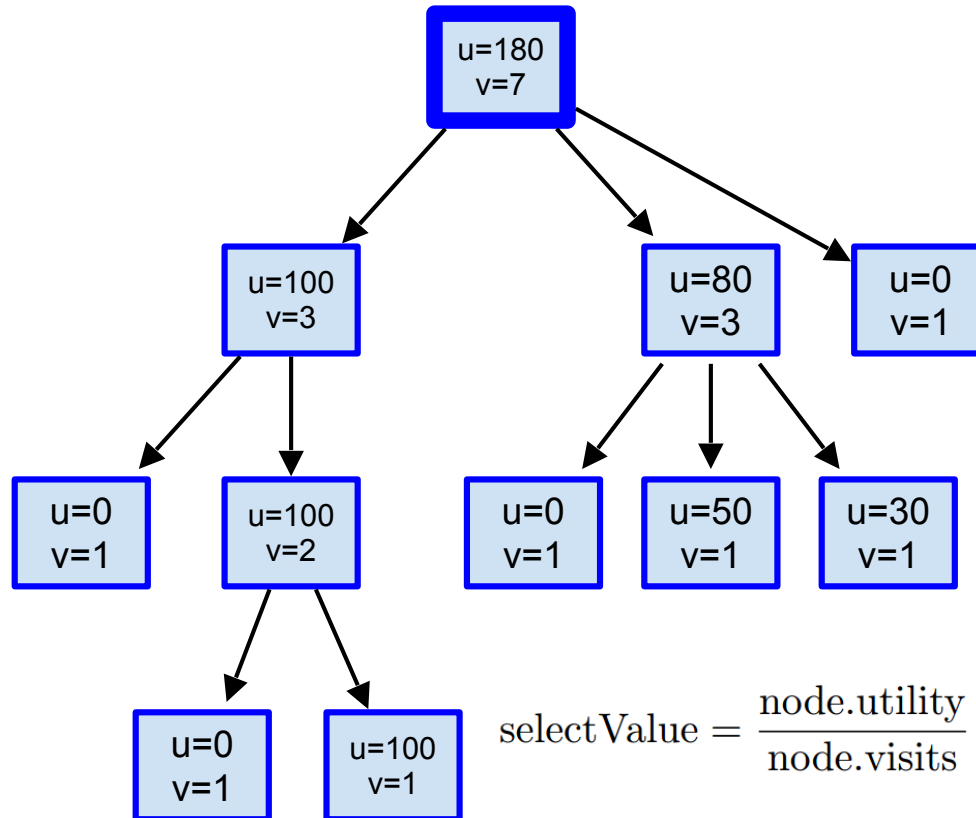
Select



Select

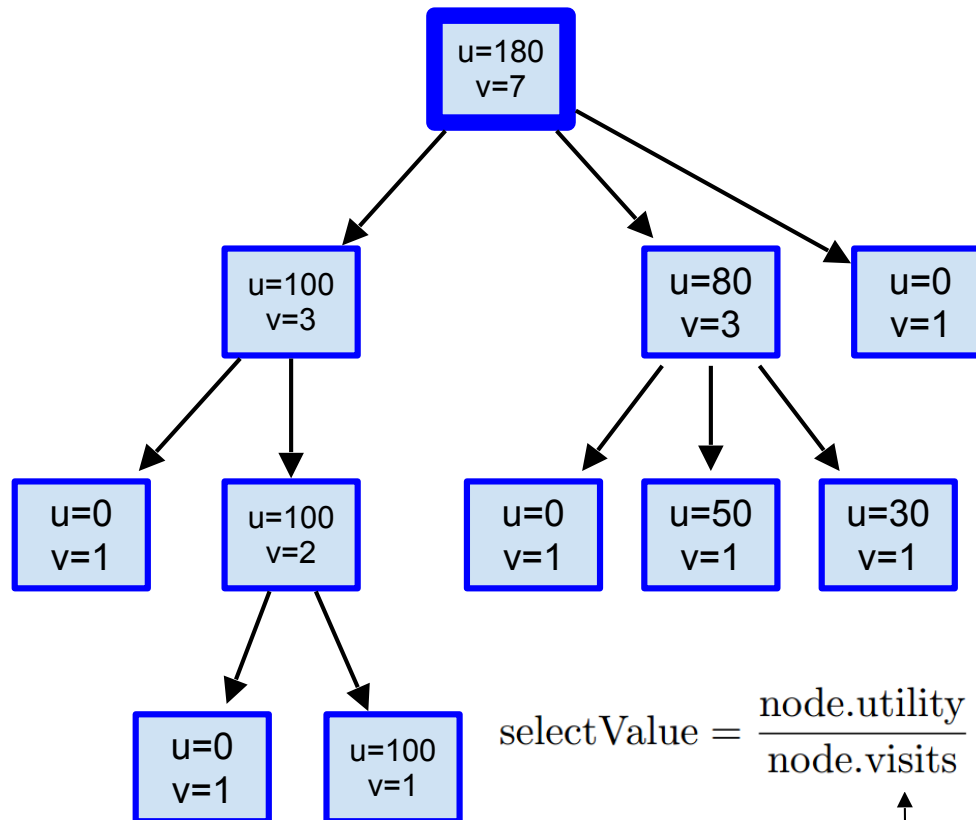


Select



$$\text{selectValue} = \frac{\text{node.utility}}{\text{node.visits}} + C \times \sqrt{\frac{\ln(\text{node.parent.visits})}{\text{node.visits}}}$$

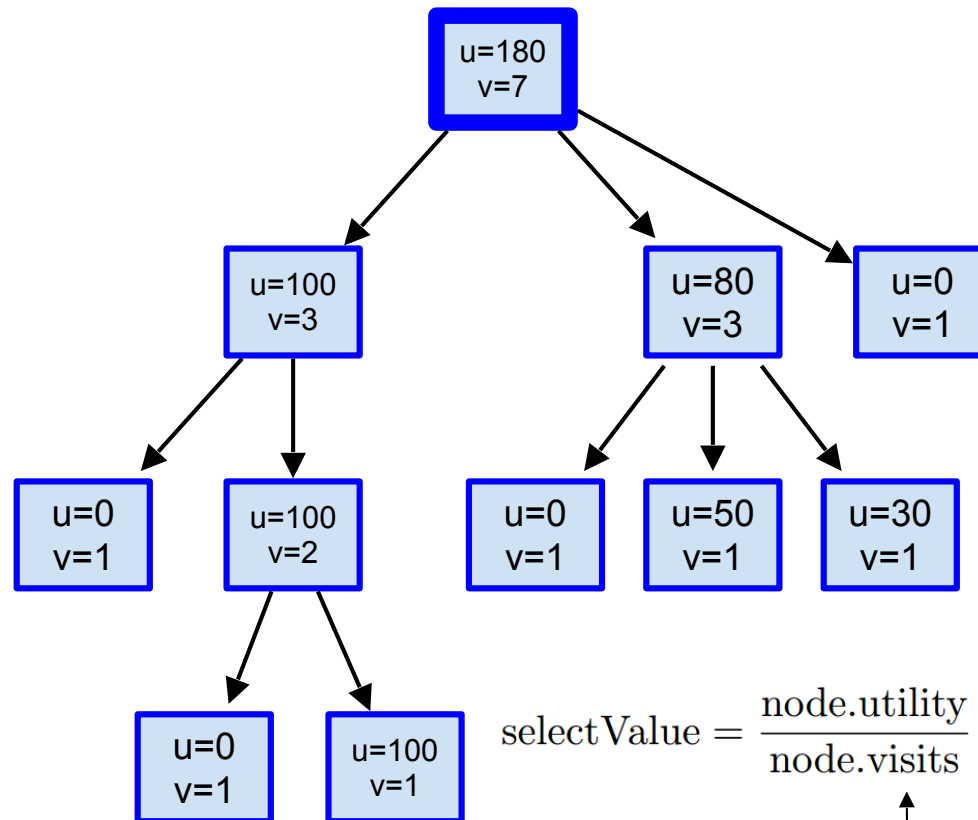
Select



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Exploitation

Select

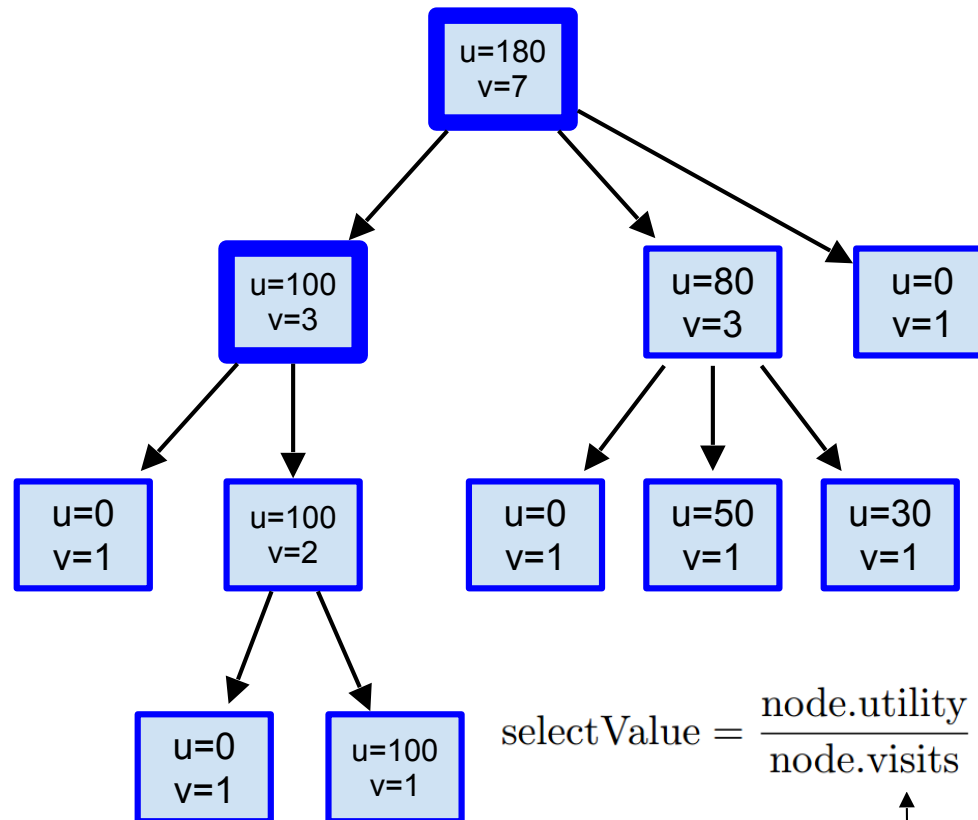


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Exploitation

Exploration

Select

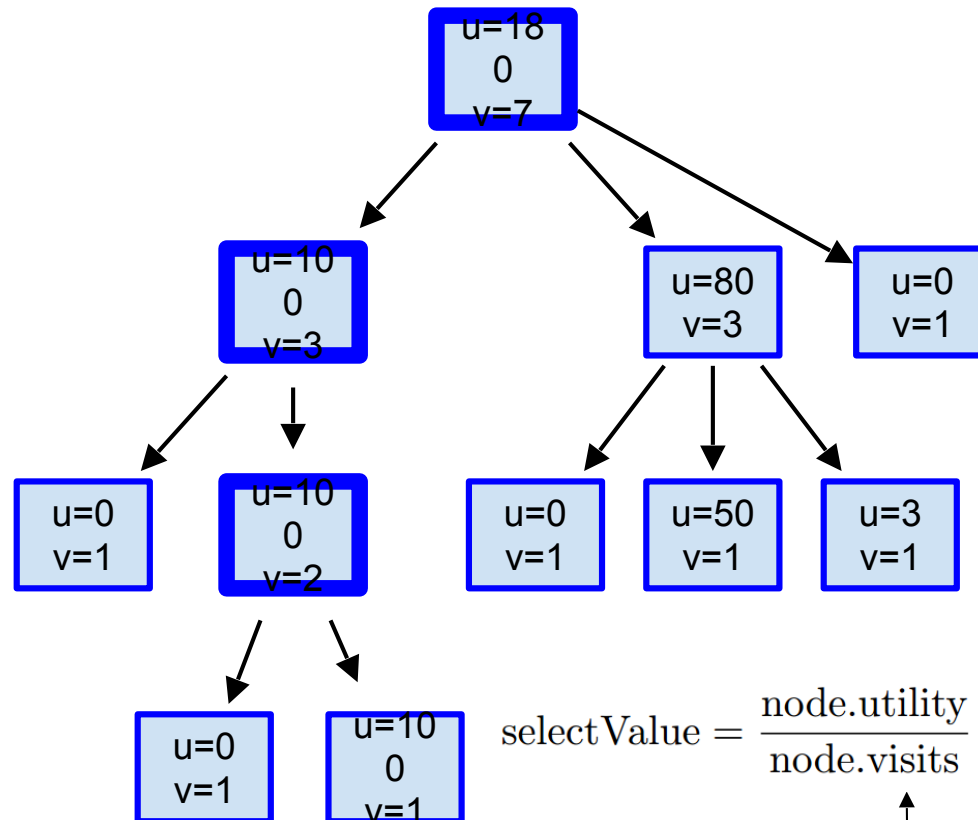


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Exploitation

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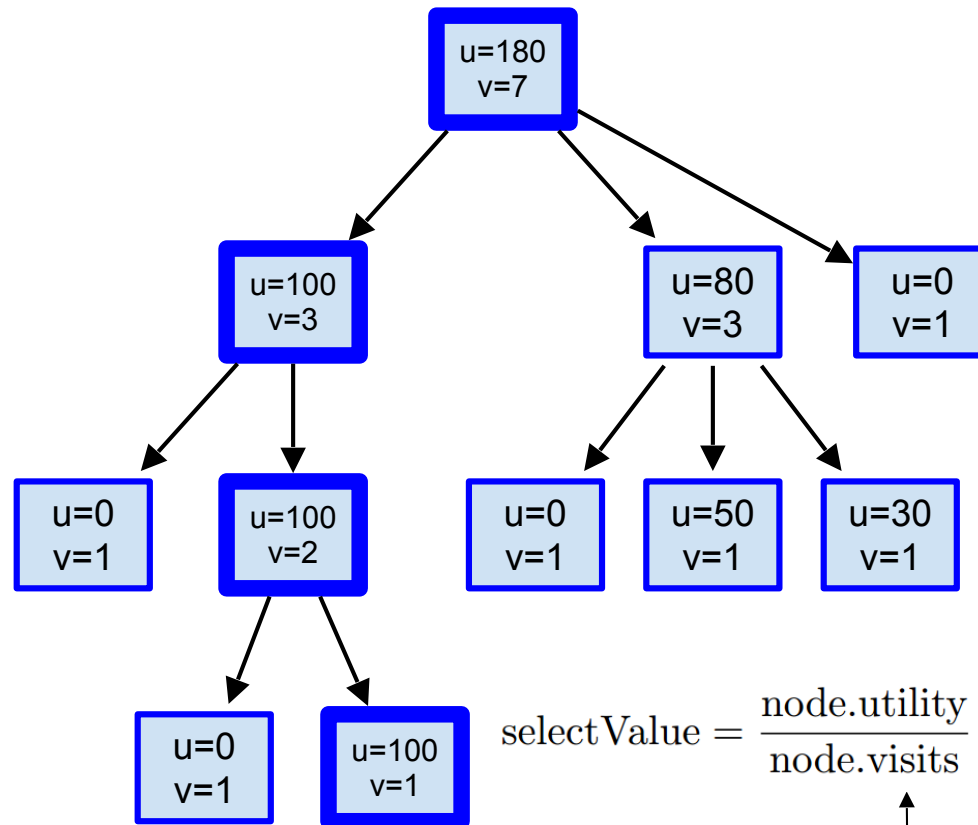


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Exploitation

Exploration

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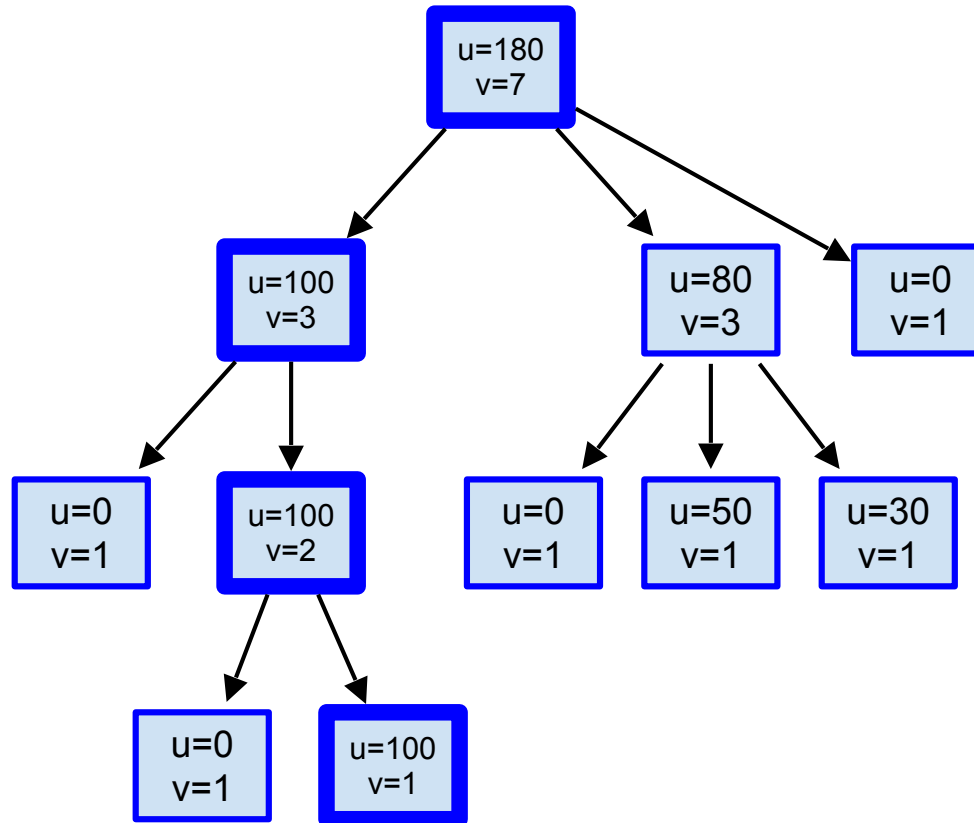


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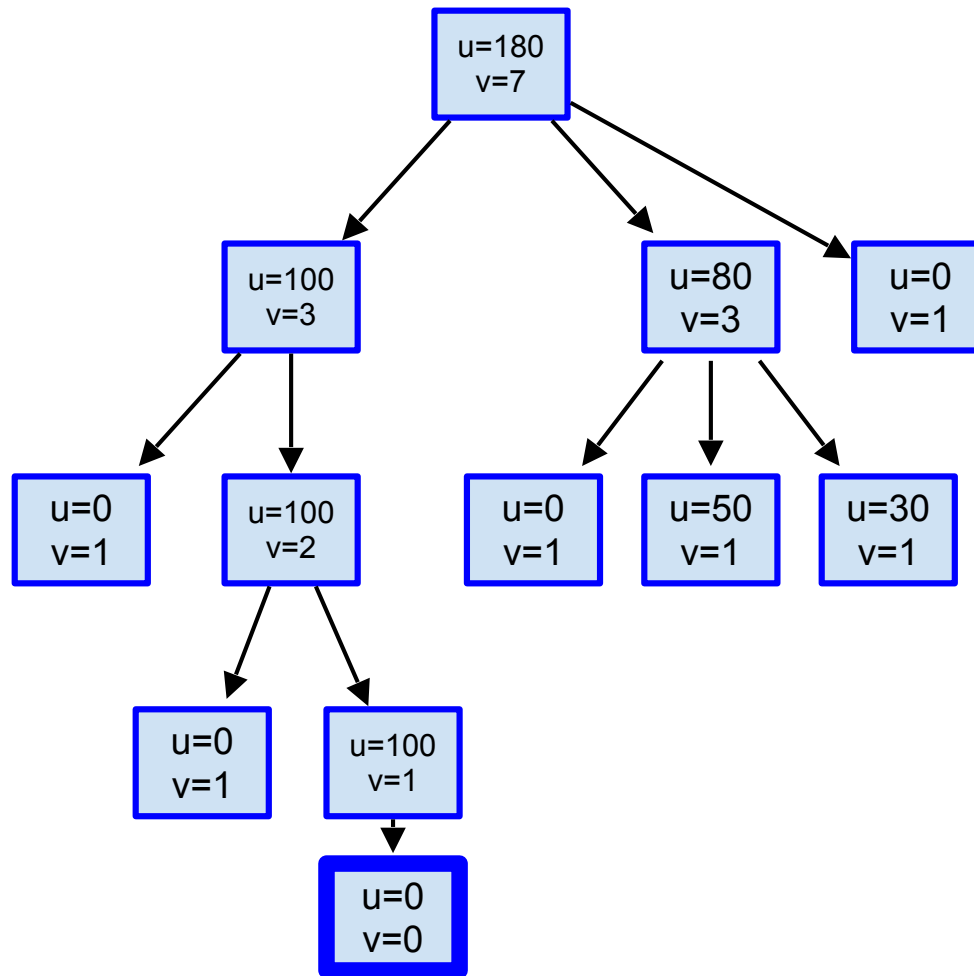
Exploitation

Exploration

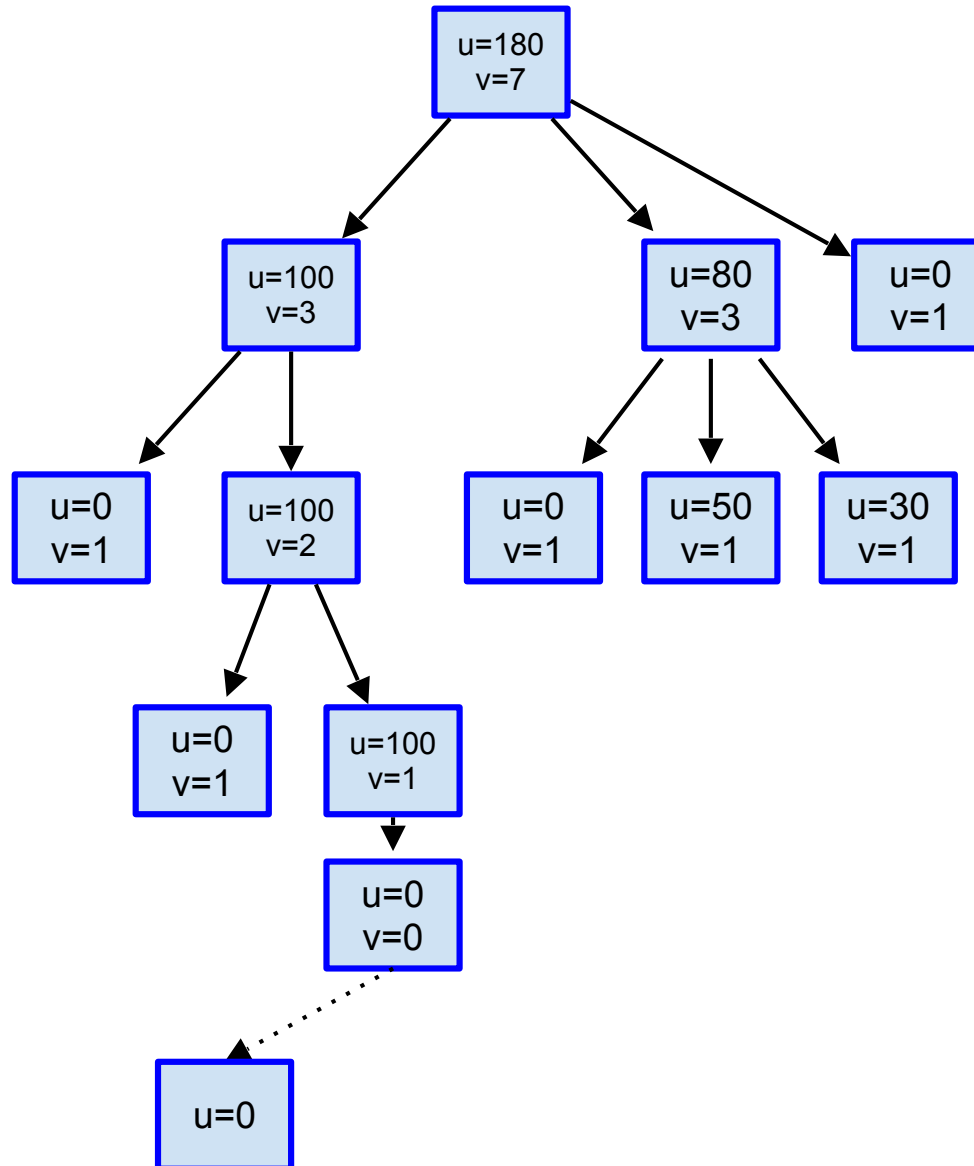
Select



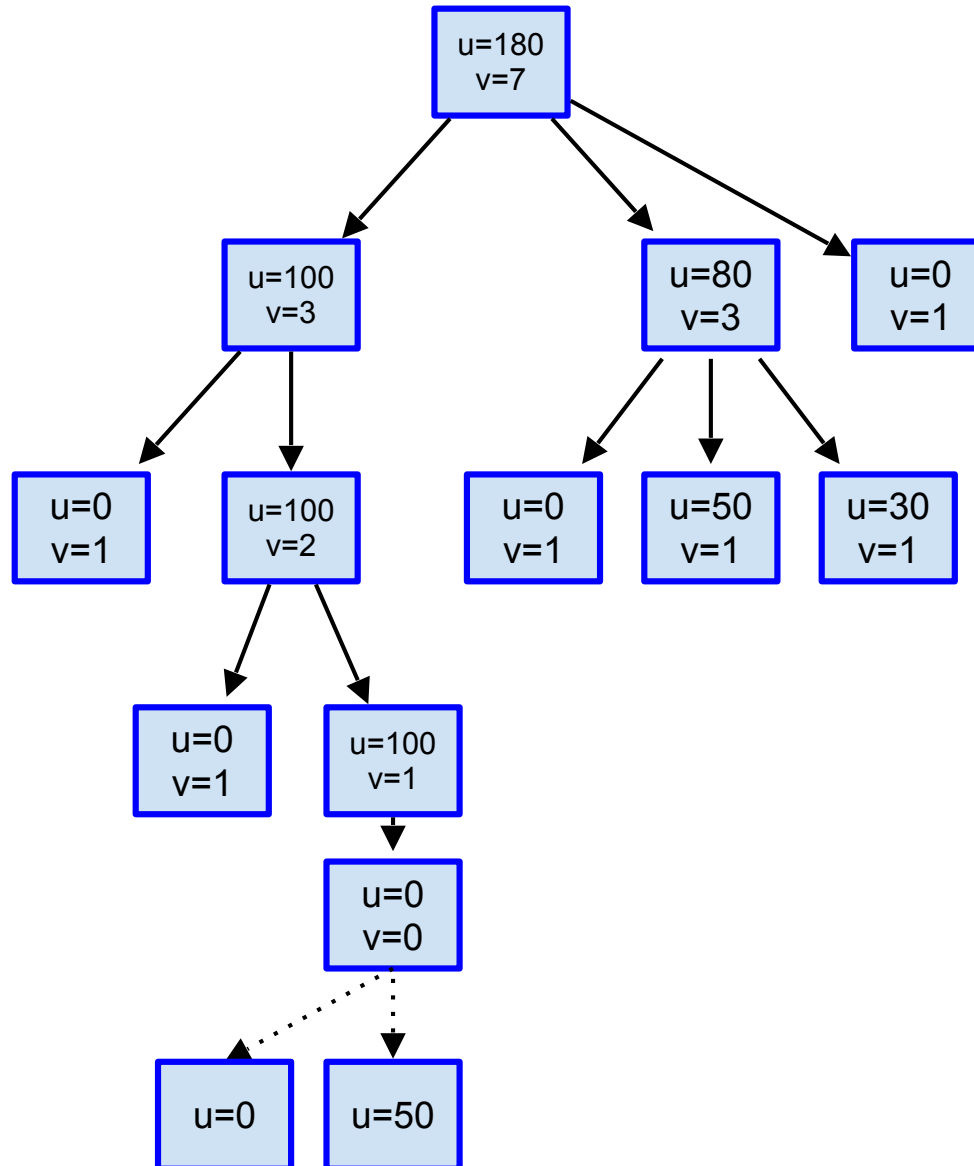
Expand



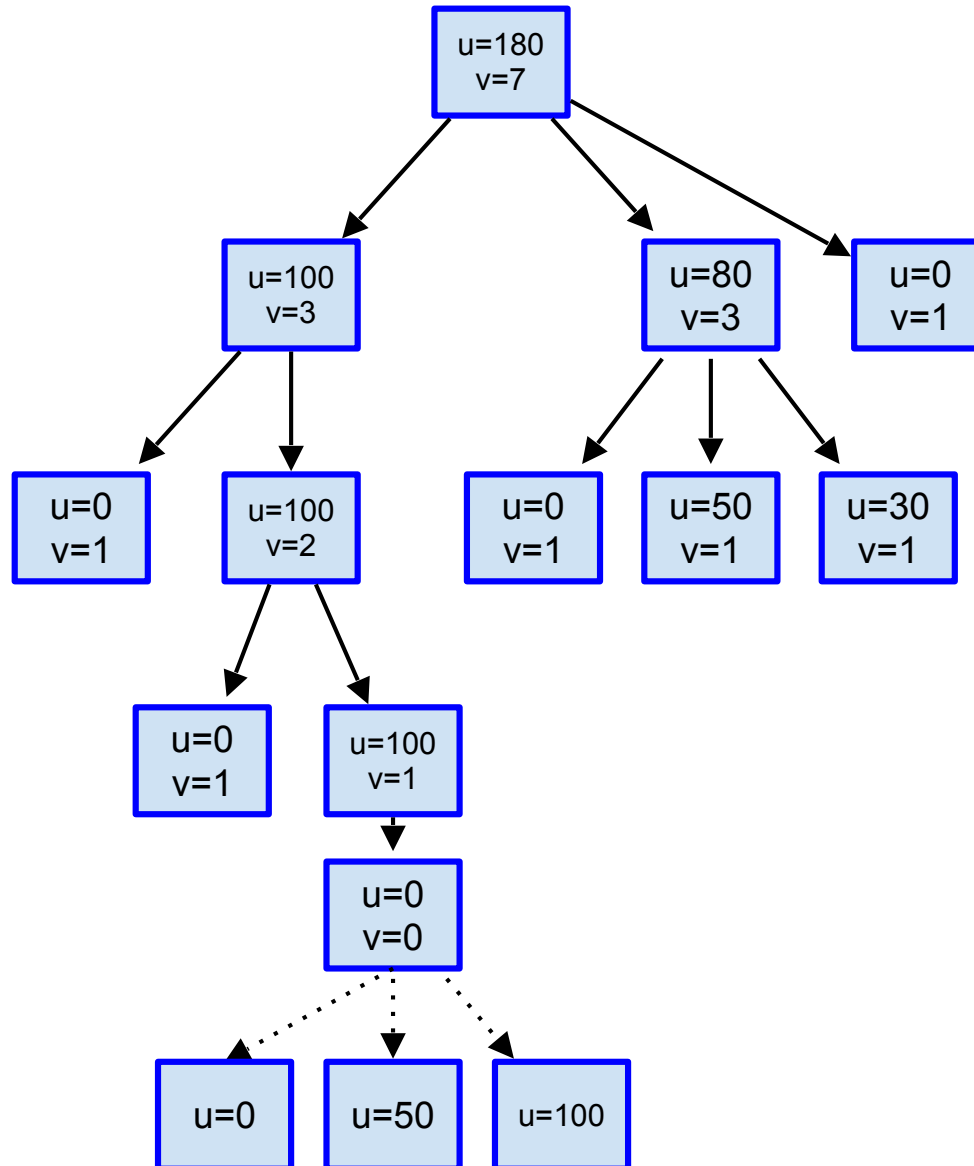
Simulate



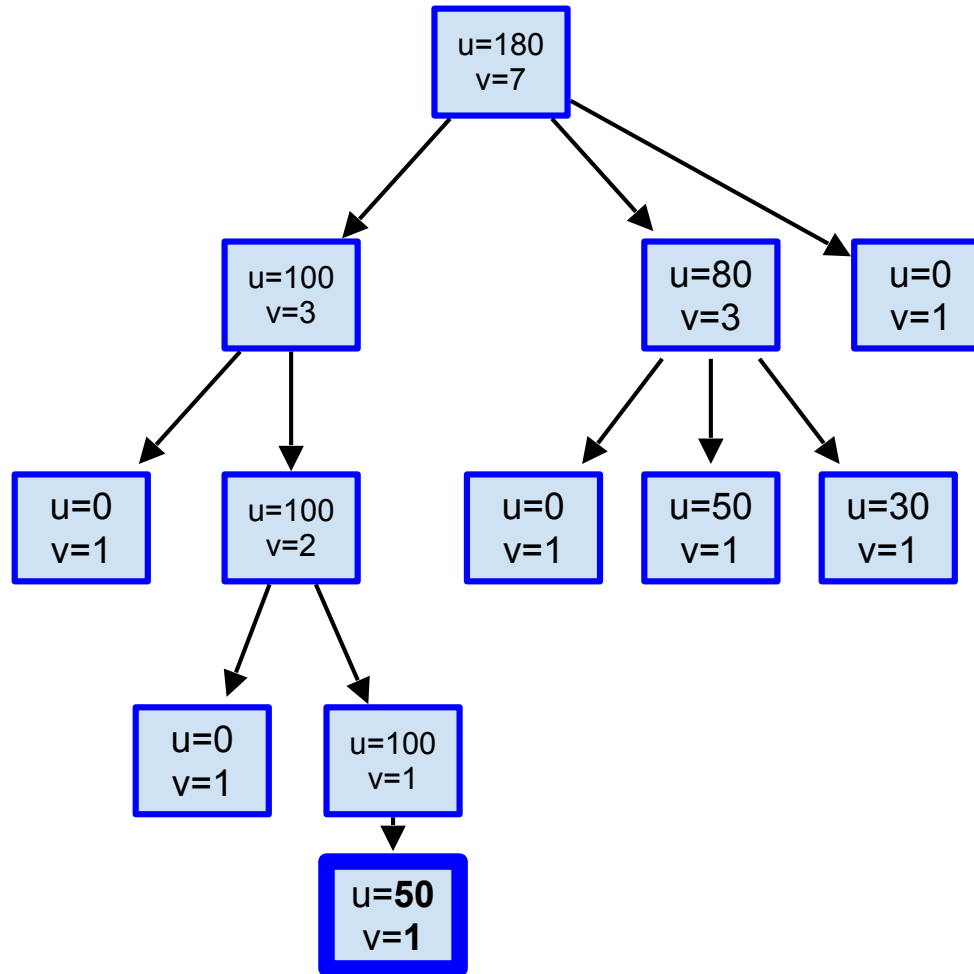
Simulate



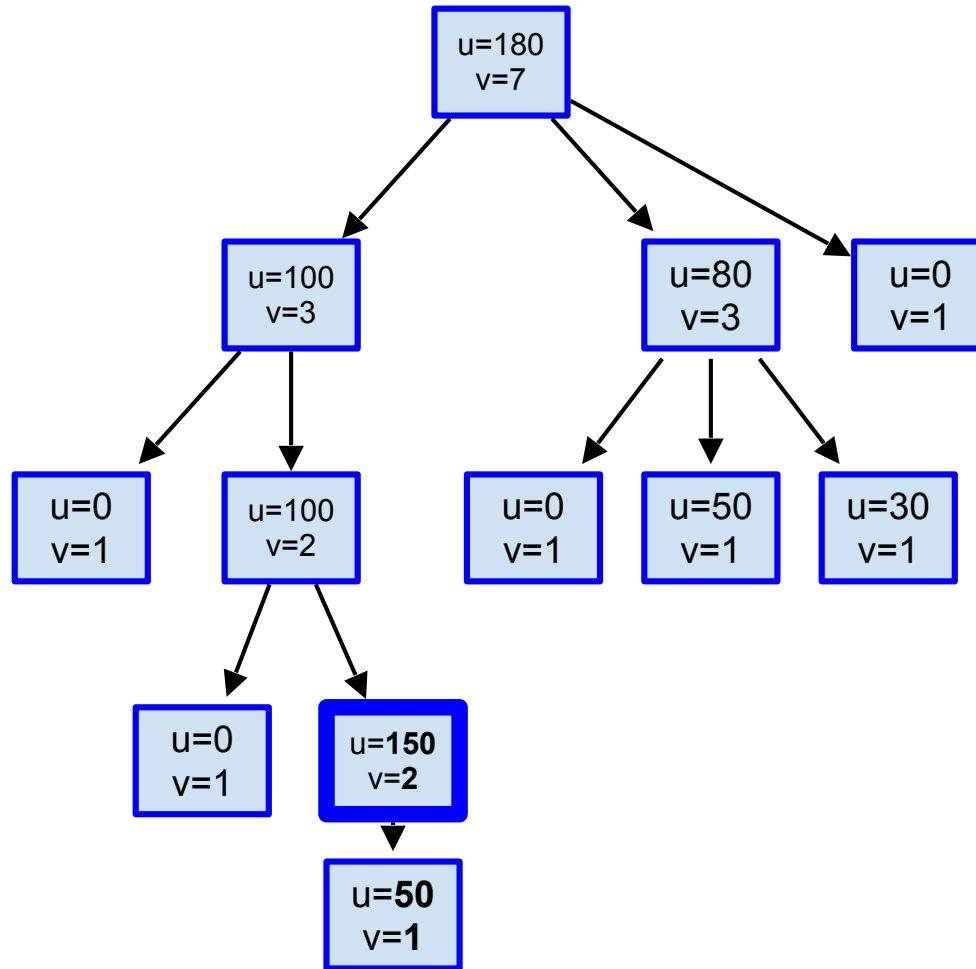
Simulate



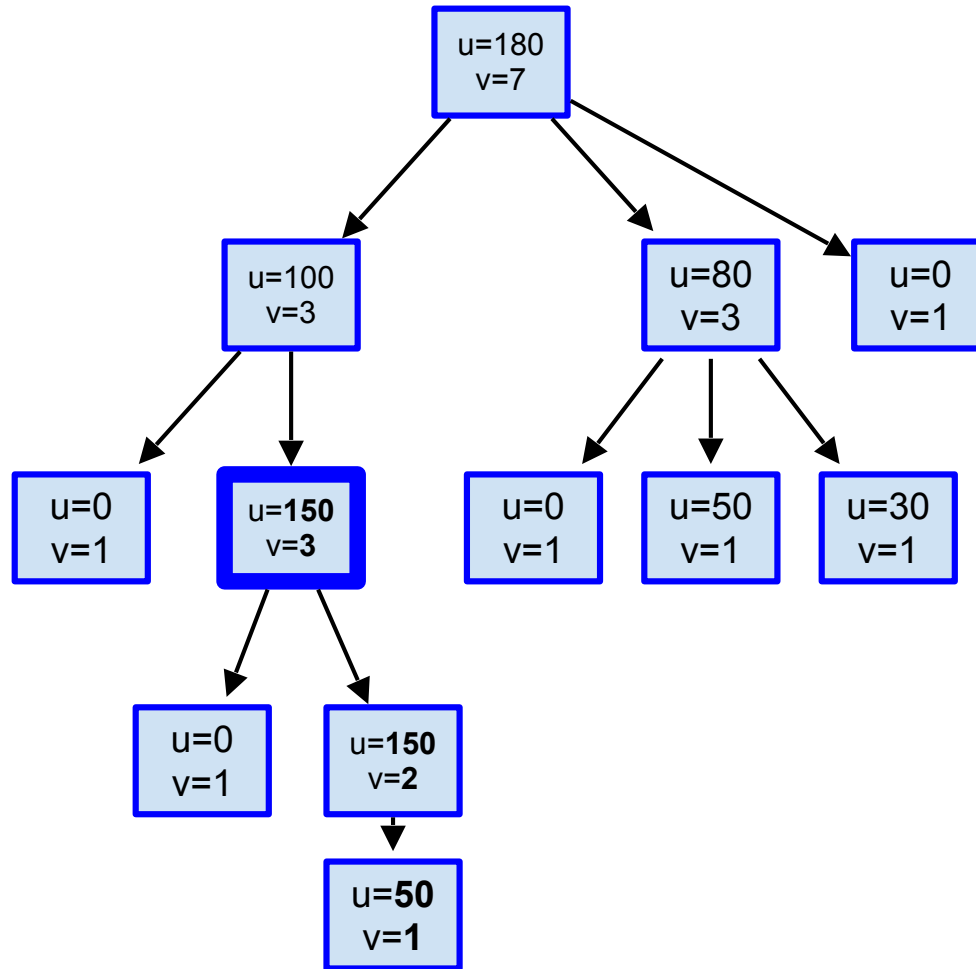
Backpropagate



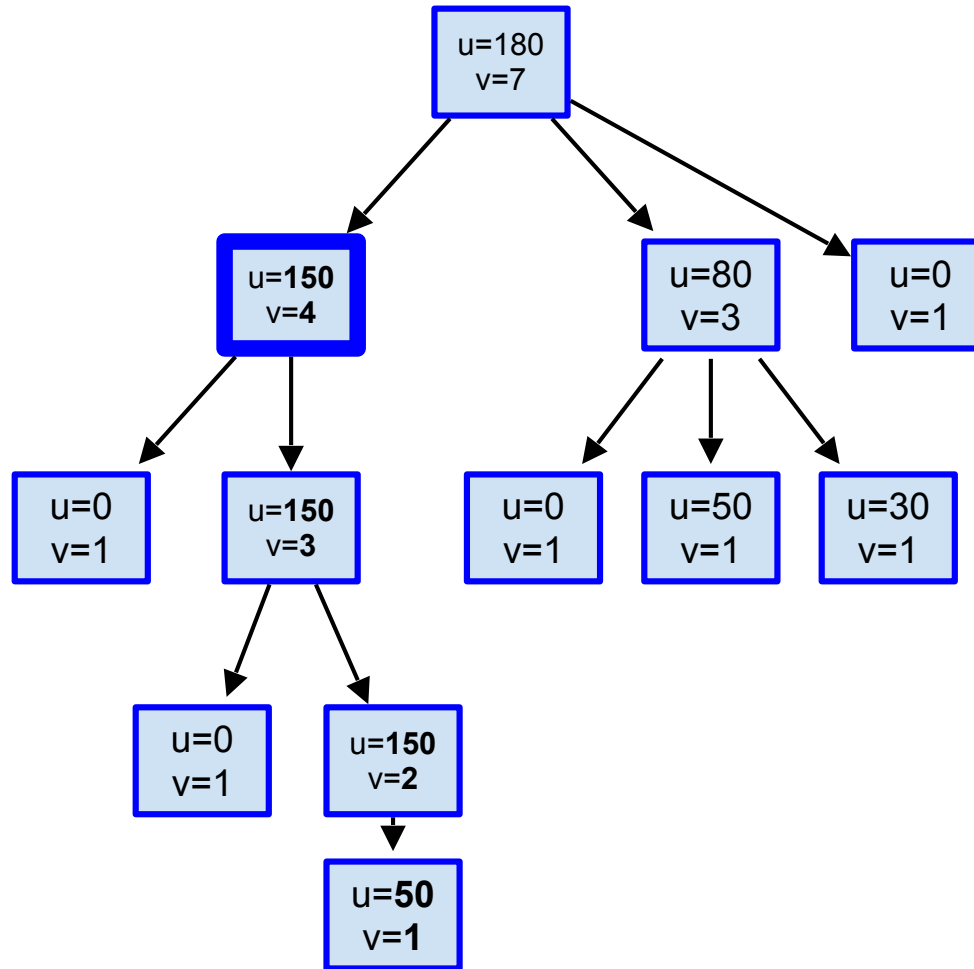
Backpropagate



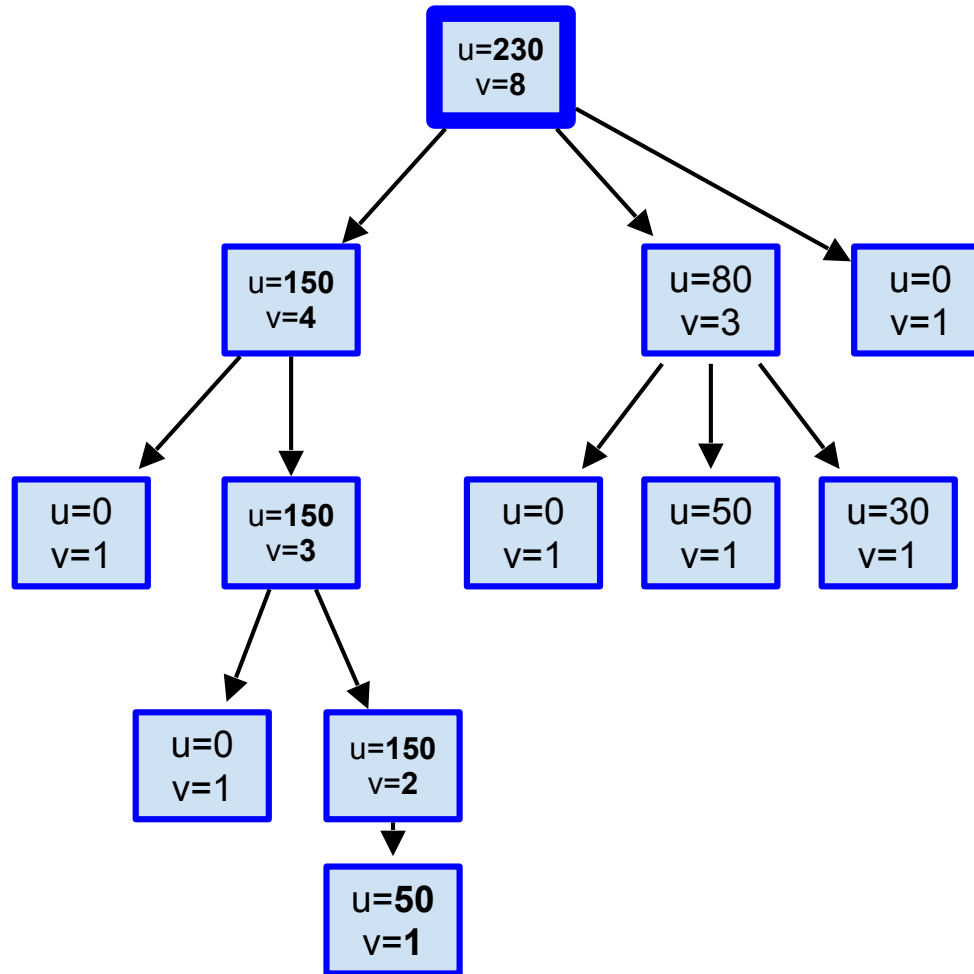
Backpropagate



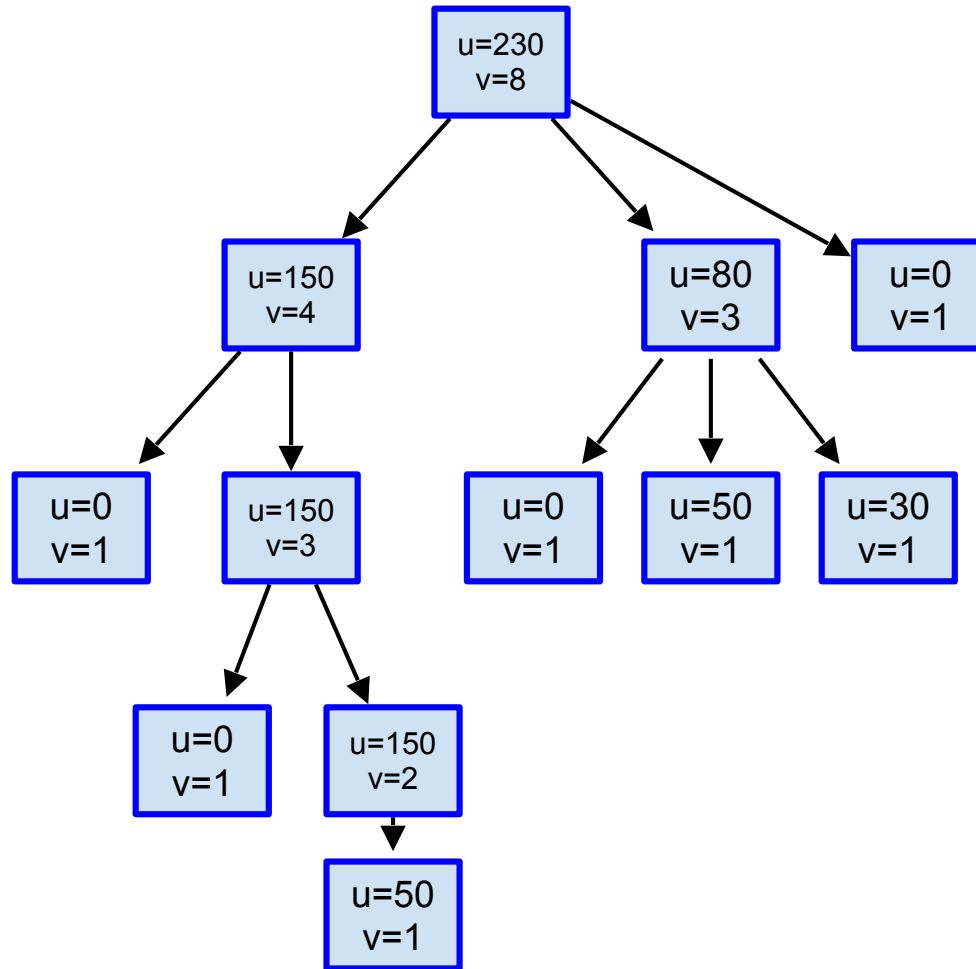
Backpropagate



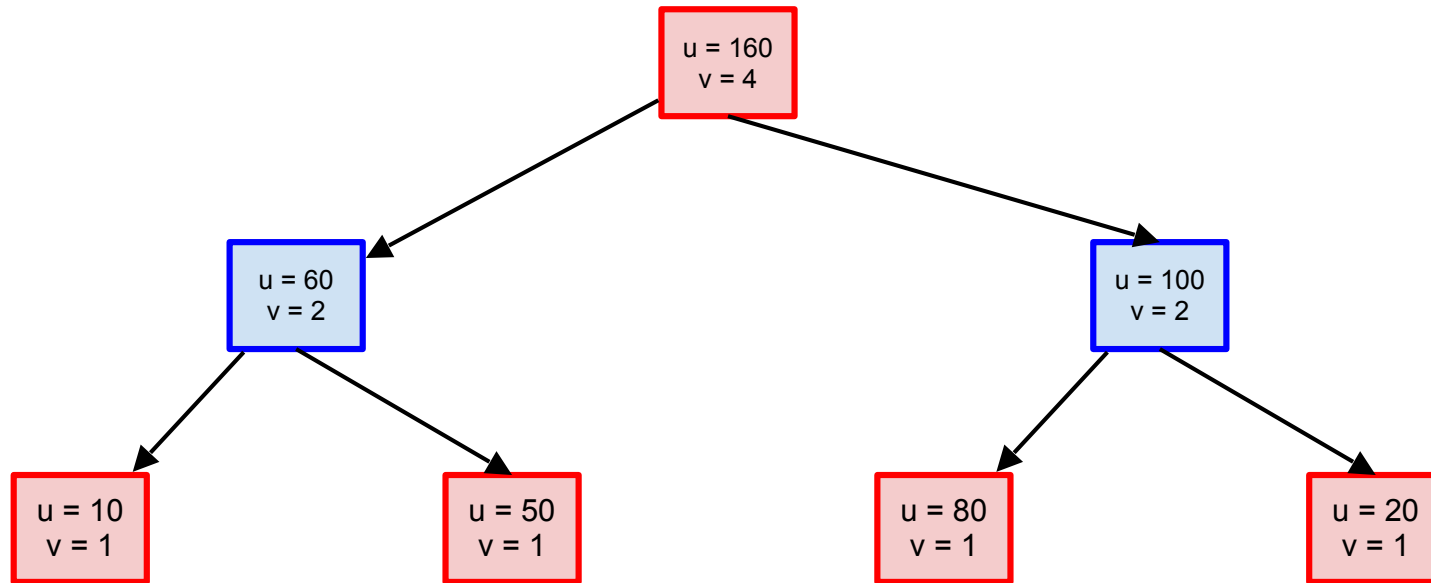
Backpropagate



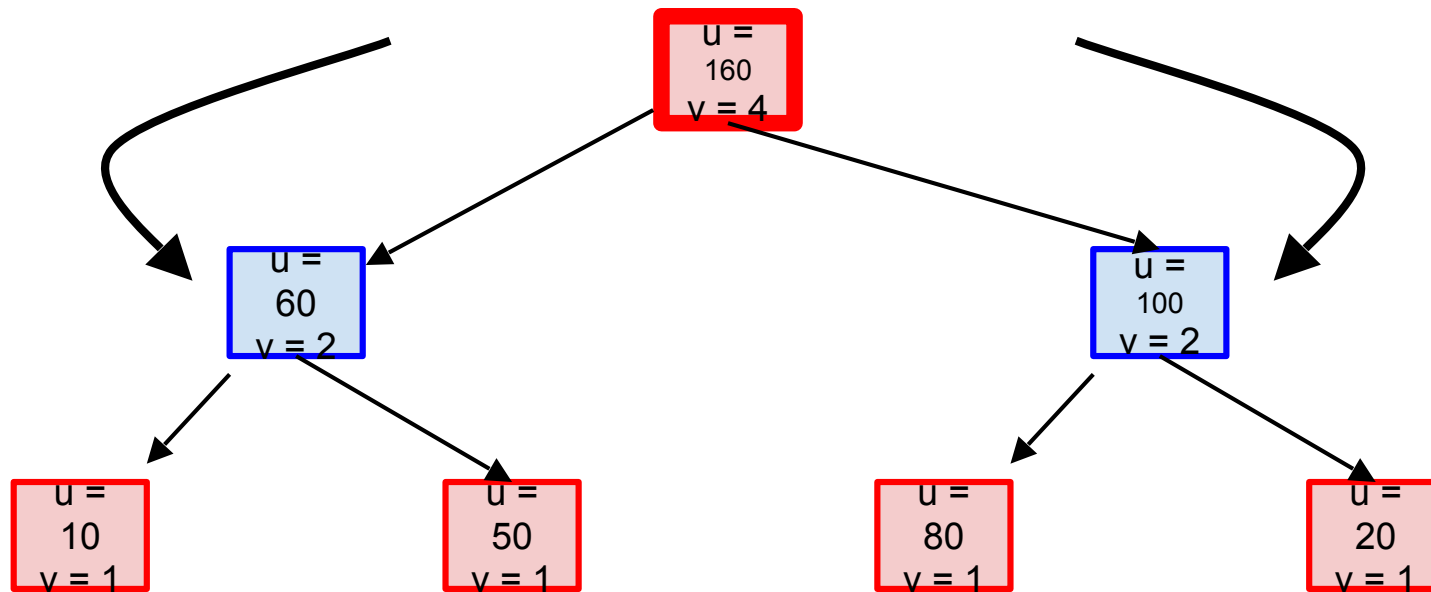
Backpropagate



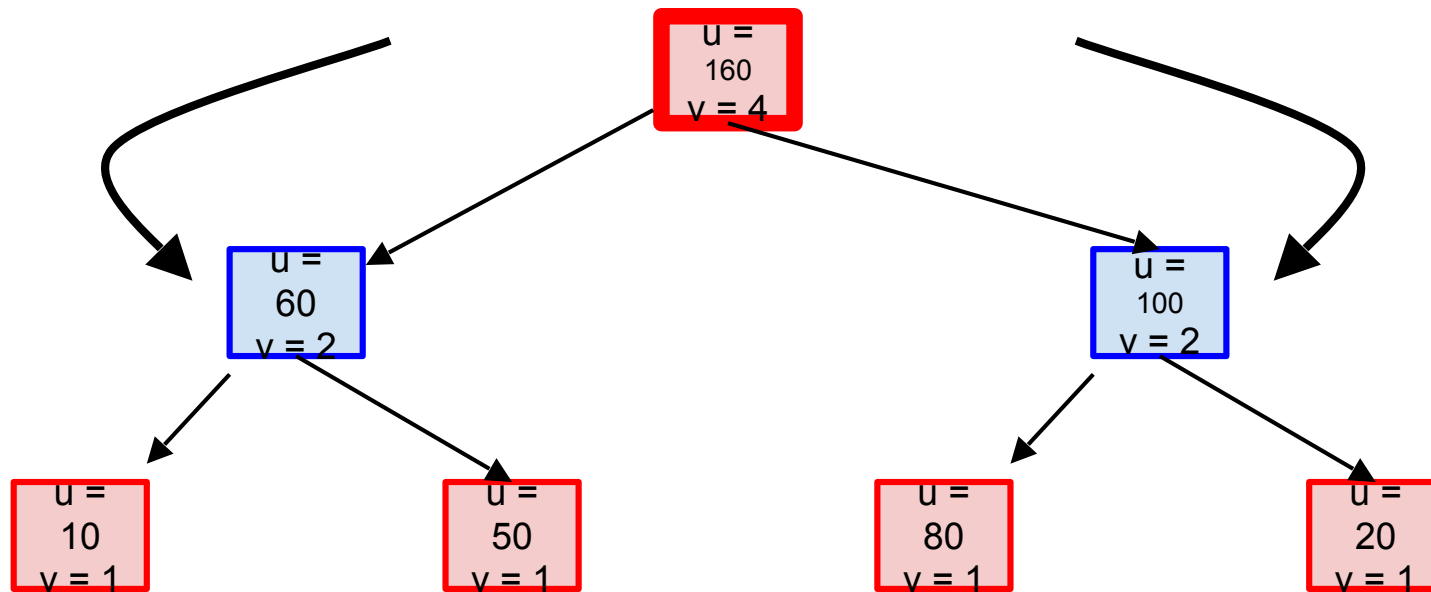
Multiple Players



Multiple Players

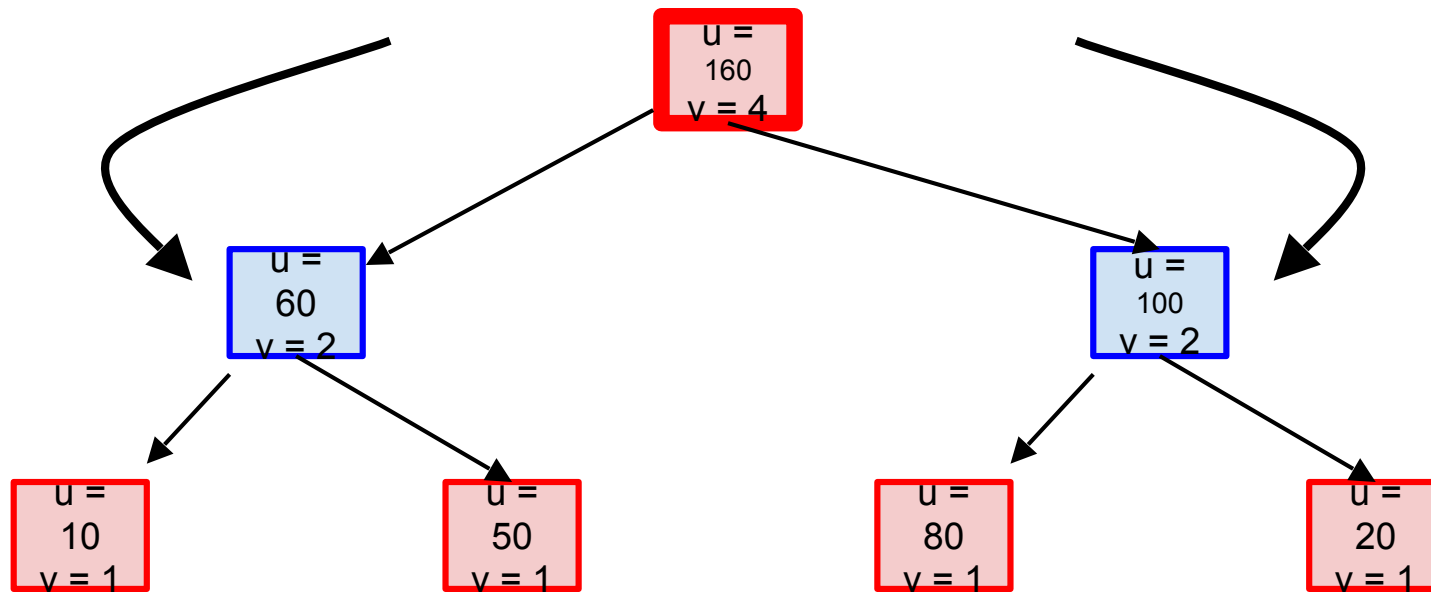


Multiple Players



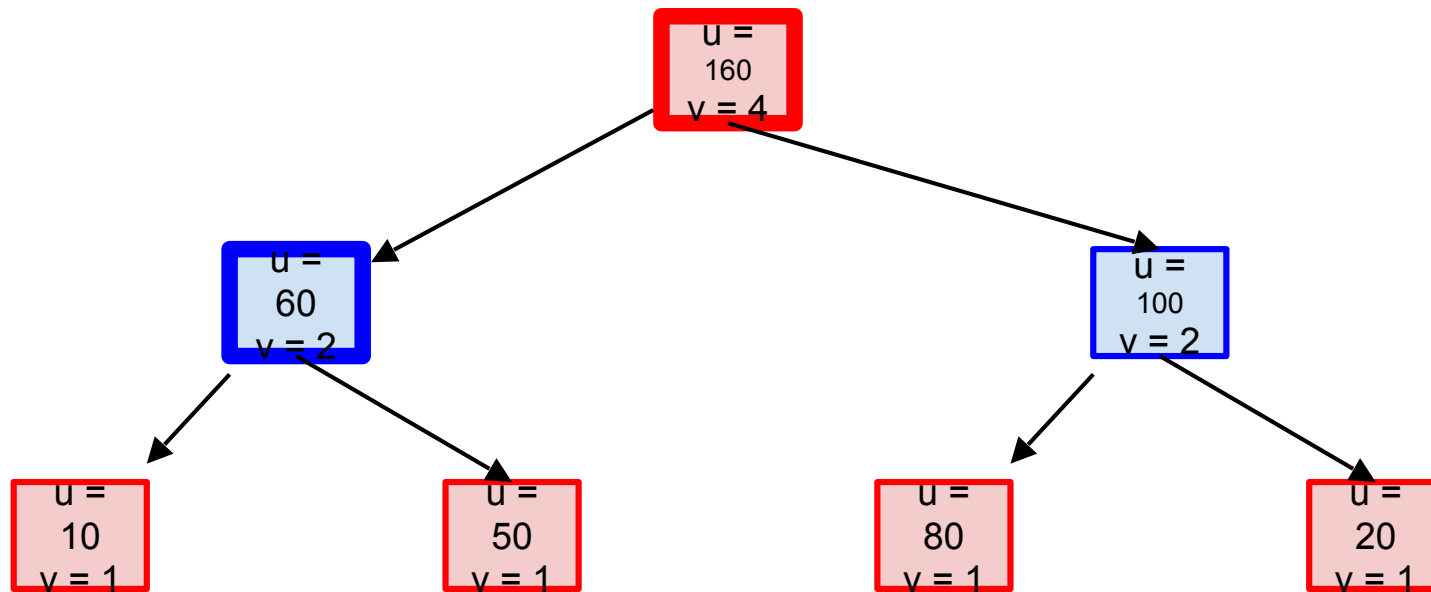
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Multiple Players

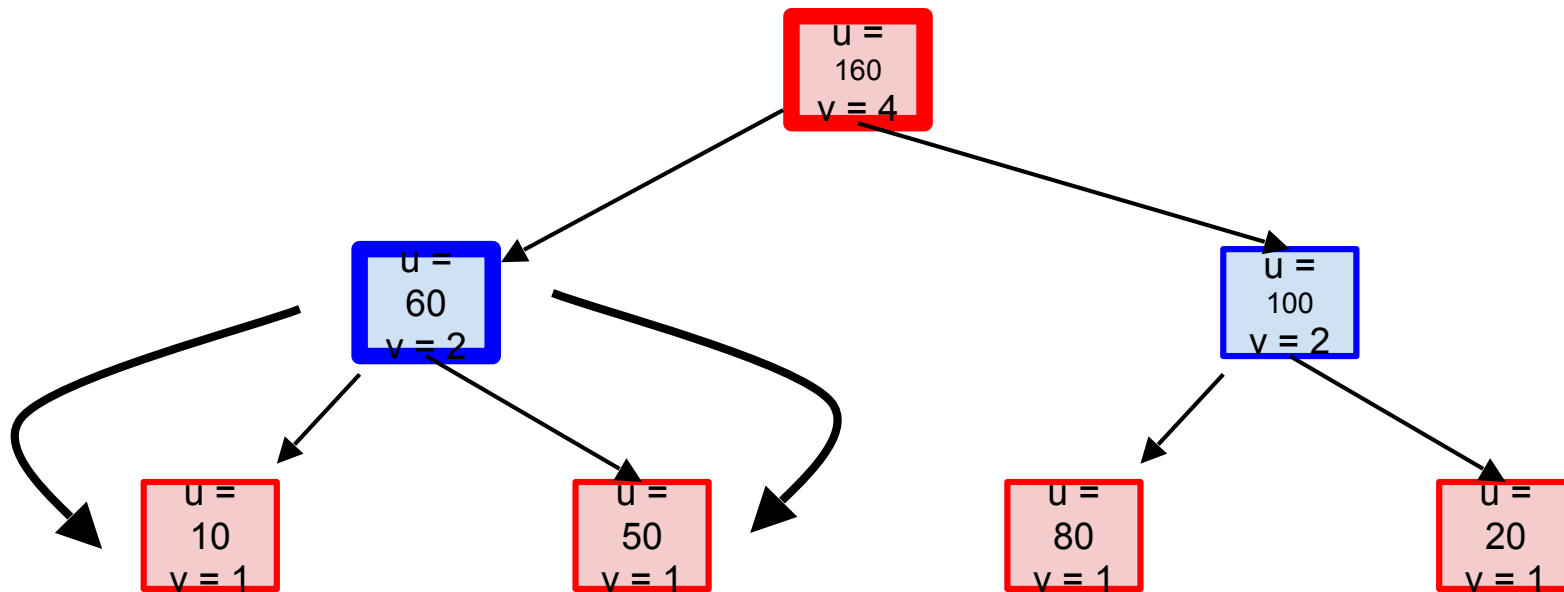


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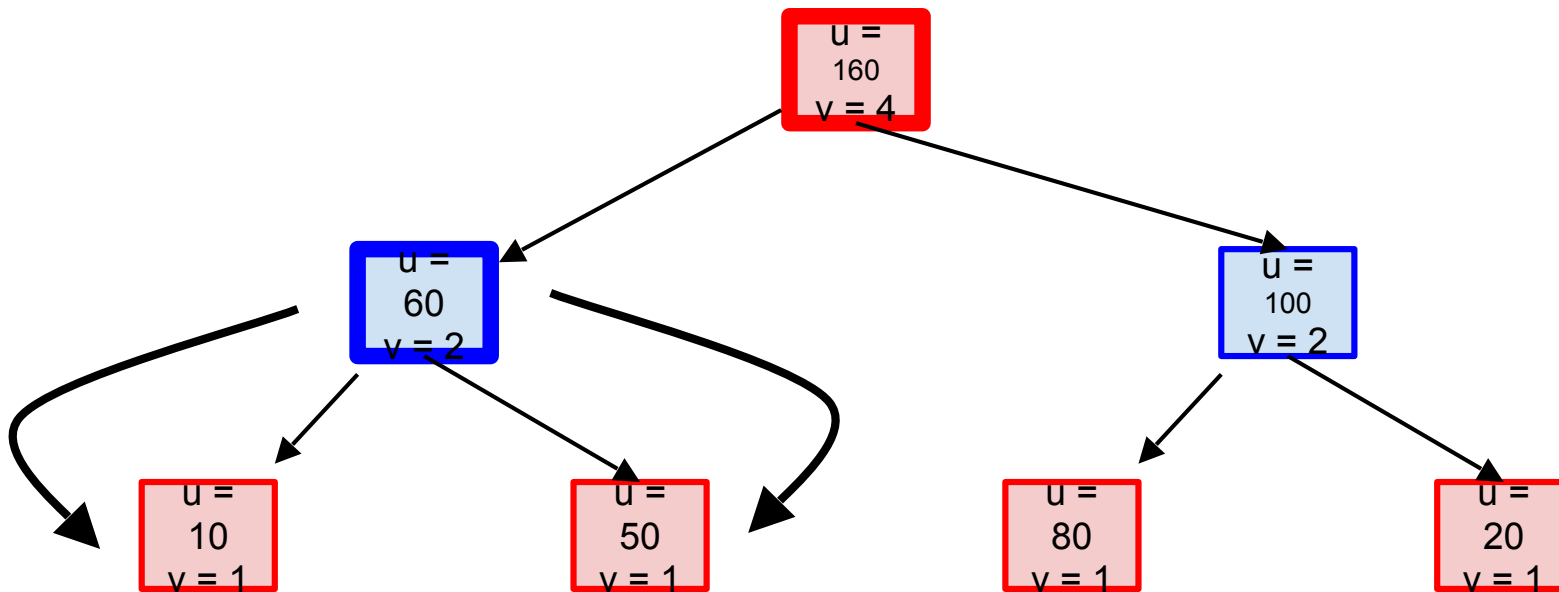
Multiple Players



Multiple Players

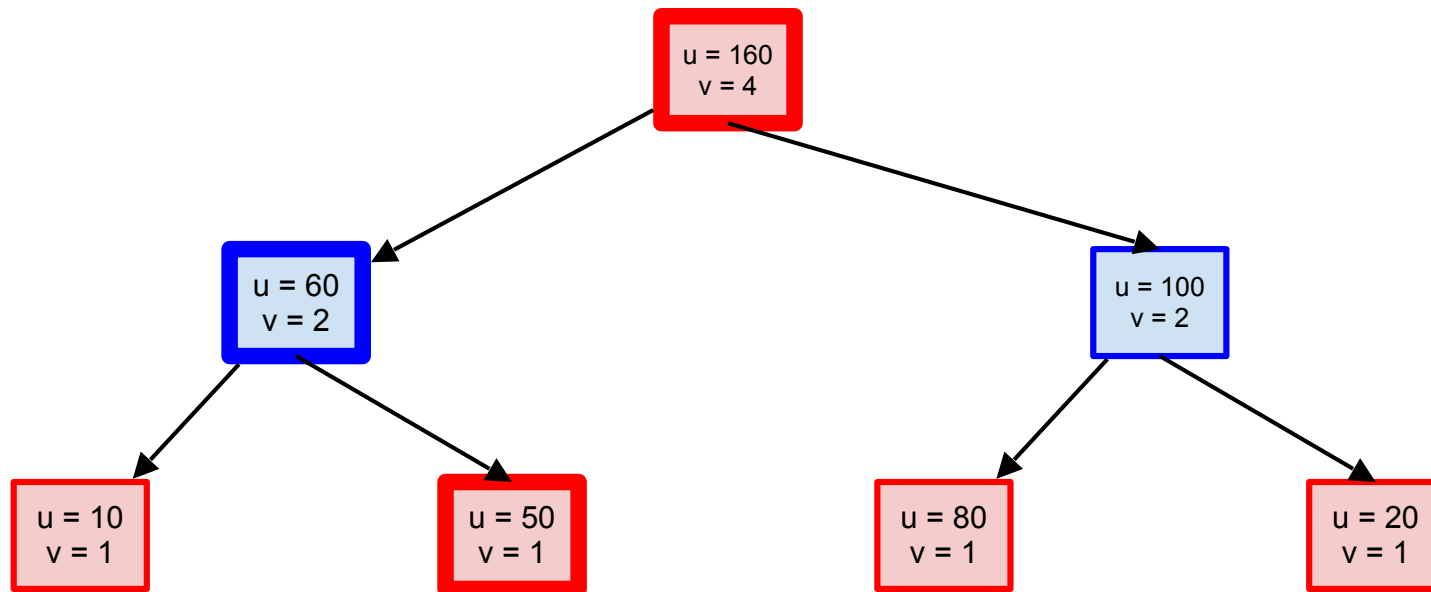


Multiple Players



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Multiple Players



Why is MCTS better than MCS?

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Move	Depth Charges
A	0, 0, 0, 0, 0, 0
B	60, 60, 60, 60, 60, 60
C	0, 100, 0, 100, 0, 100

Why is MCTS better than MCS?

Move	Depth Charges	Mean
A	0, 0, 0, 0, 0, 0	0
B	60, 60, 60, 60, 60, 60	60
C	0, 100, 0, 100, 0, 100	50

Why is MCTS better than MCS?

Move	Depth Charges	Mean	Std dev
A	0, 0, 0, 0, 0, 0	0	0
B	60, 60, 60, 60, 60, 60	60	0
C	0, 100, 0, 100, 0, 100	50	54.8

More Information

https://en.wikipedia.org/wiki/Monte_Carlo_tree_search



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