

DATA130008 Introduction to Artificial Intelligence

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Gomoku Tutorial 1

November 02th 2021

- Gomoku Rule
- Solve Gomoku
 - Board Representation
 - Monte Carlo Tree Search
 - Proof-Number Search
 - Threat-Space Search
 - Genetic Algorithm

- **Gomoku Rule**
- Solve Gomoku
 - Board Representation
 - Monte Carlo Tree Search
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- **Goal: Five in a row, 15×15**
 - **Proved: Black first leads to win (1899)**
- **Gomoku without forbidden shape:**
 - **Free: 5 or more than 5**
 - **Standard: only 5**
 - **Swap 2 Rule**
- **Renju: forbid some shape for Black**

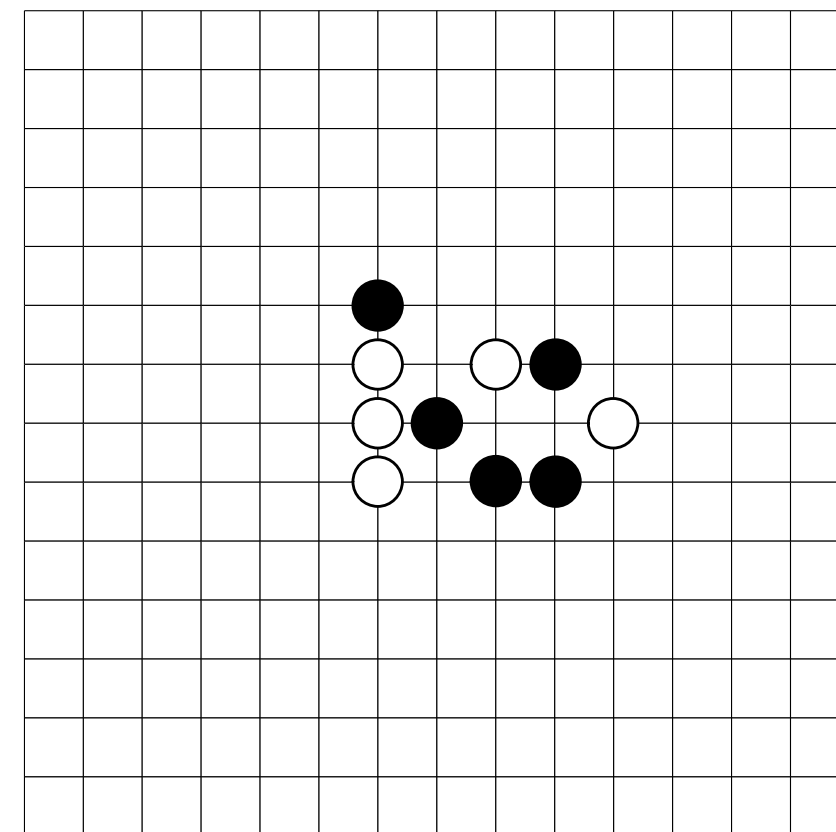
- **Focus on: Free Gomoku**

- Gomoku Rule
- **Solve Gomoku**
 - Board Representation
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- Input
 - Current board state
- Goal
 - Search for next step

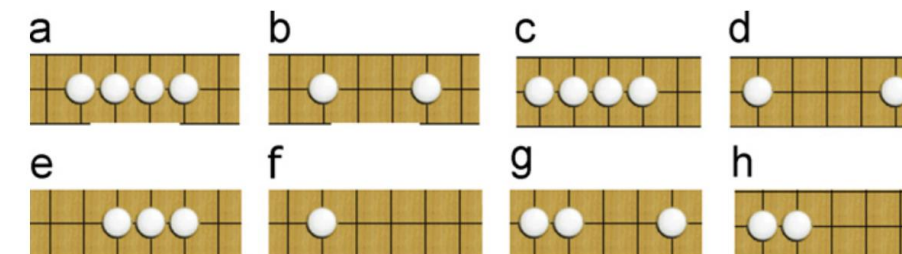
- Gomoku Rule
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 - **Board Representation**
 - Monte Carlo Tree Search
 - Proof-Number Search
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- Board representation
 - Atomic
 - $(x, y, \bullet / \circ / \cdot)$
 - Structured: Features
 - Patterns
 - Turns
 - Offensive/Defensive

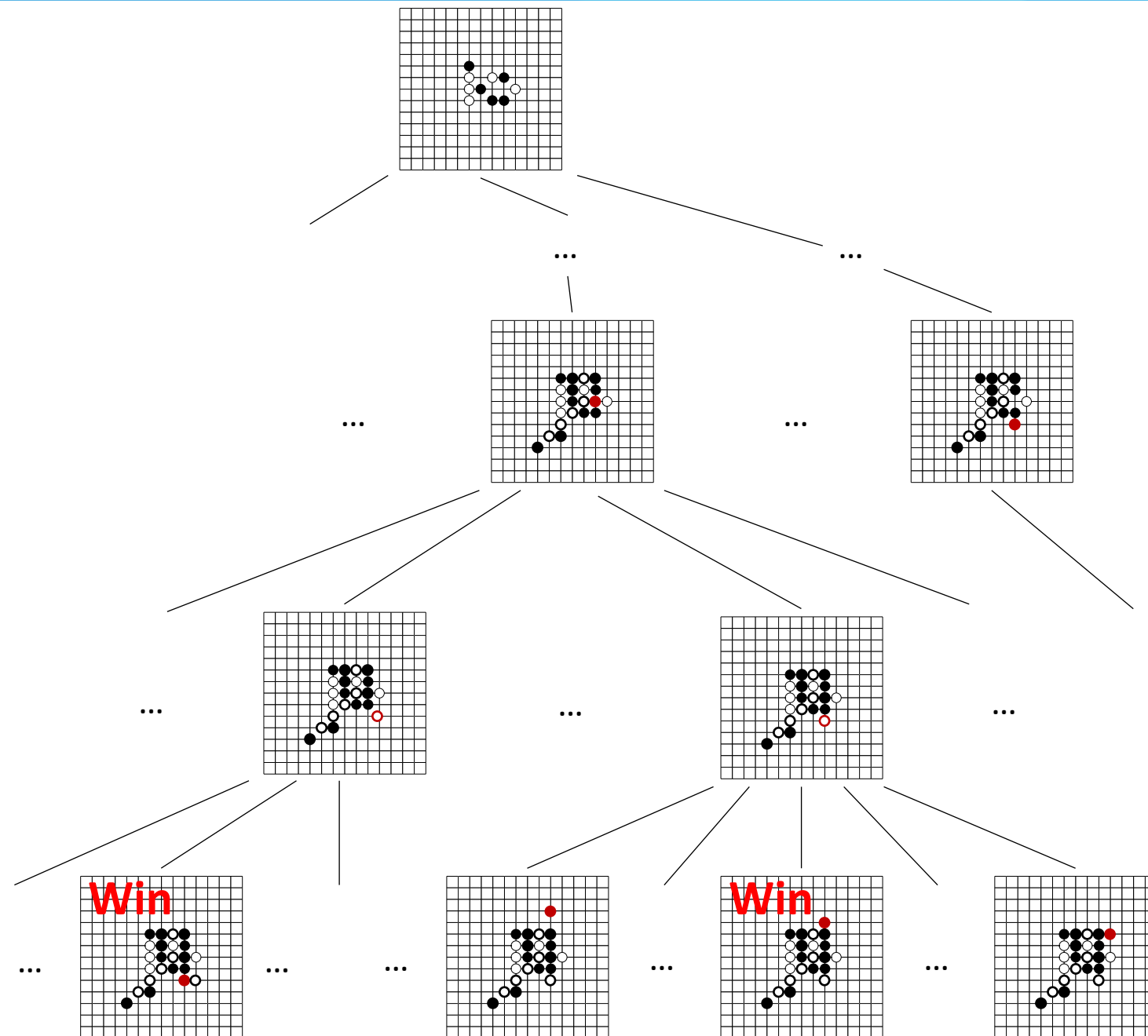


Dongbin Zhao, Zhen Zhang, and Yujie Dai.
Self-teaching adaptive dynamic programming for Gomoku.
Neurocomputing, 78(1):23–29, 2012.

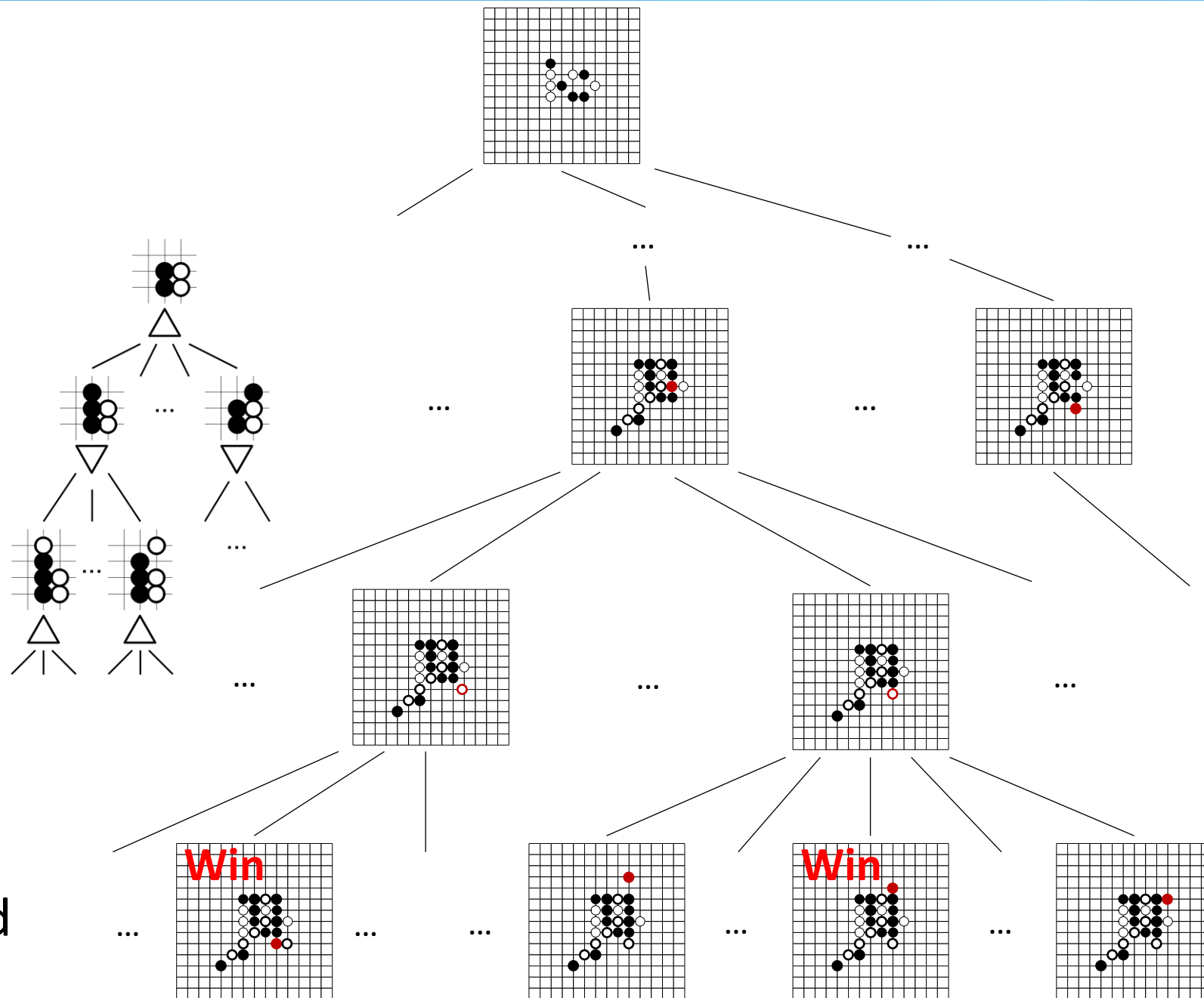
Zhentao Tang, Dongbin Zhao, Kun Shao, and Le Lv.
ADP with MCTS algorithm for Gomoku.
2016 IEEE Symp. Ser. Comput. Intell. SSCI 2016, (61273136), 2017.



- Input
 - Current board state
- Goal
 - Search for next step
 - Single agent
 - Adversarial agent
- Construct a search tree
 - Node: Gomoku board



- Input
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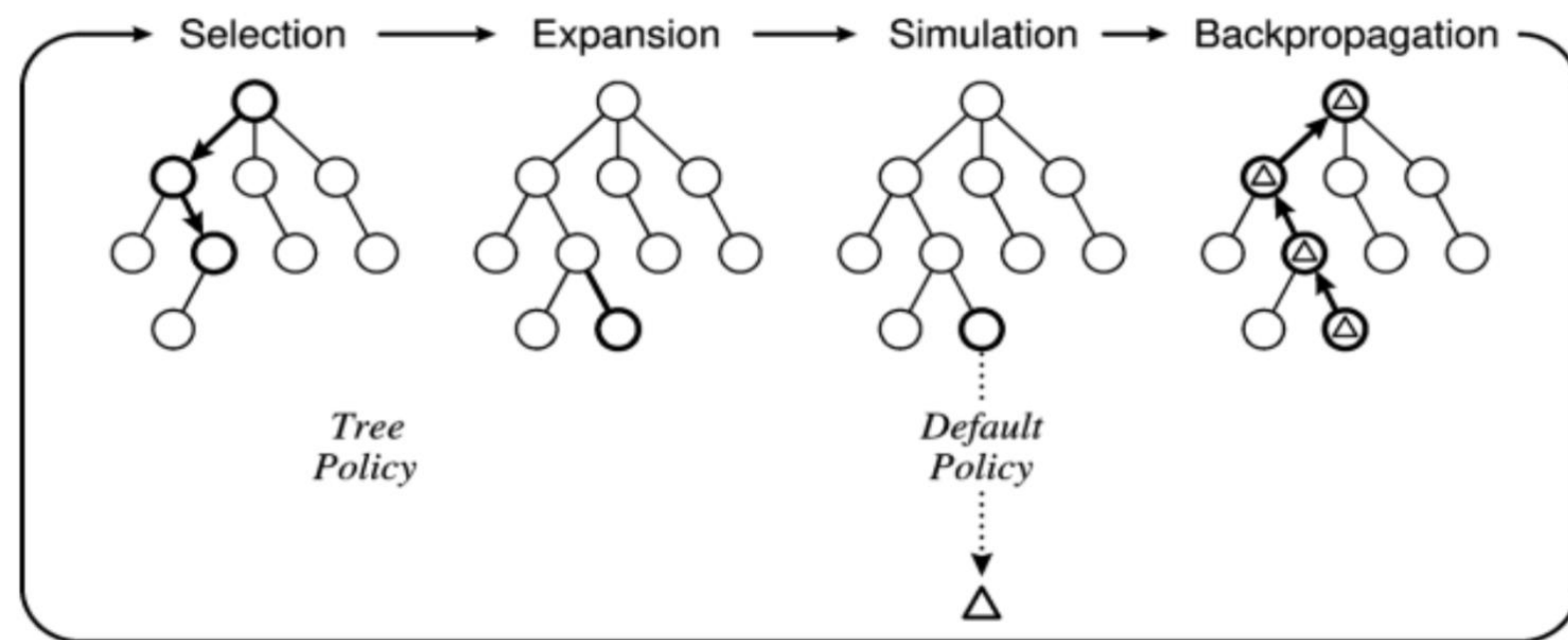
- Gomoku Rule
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 - **Monte Carlo Tree Search**
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- Monte Carlo Tree Search

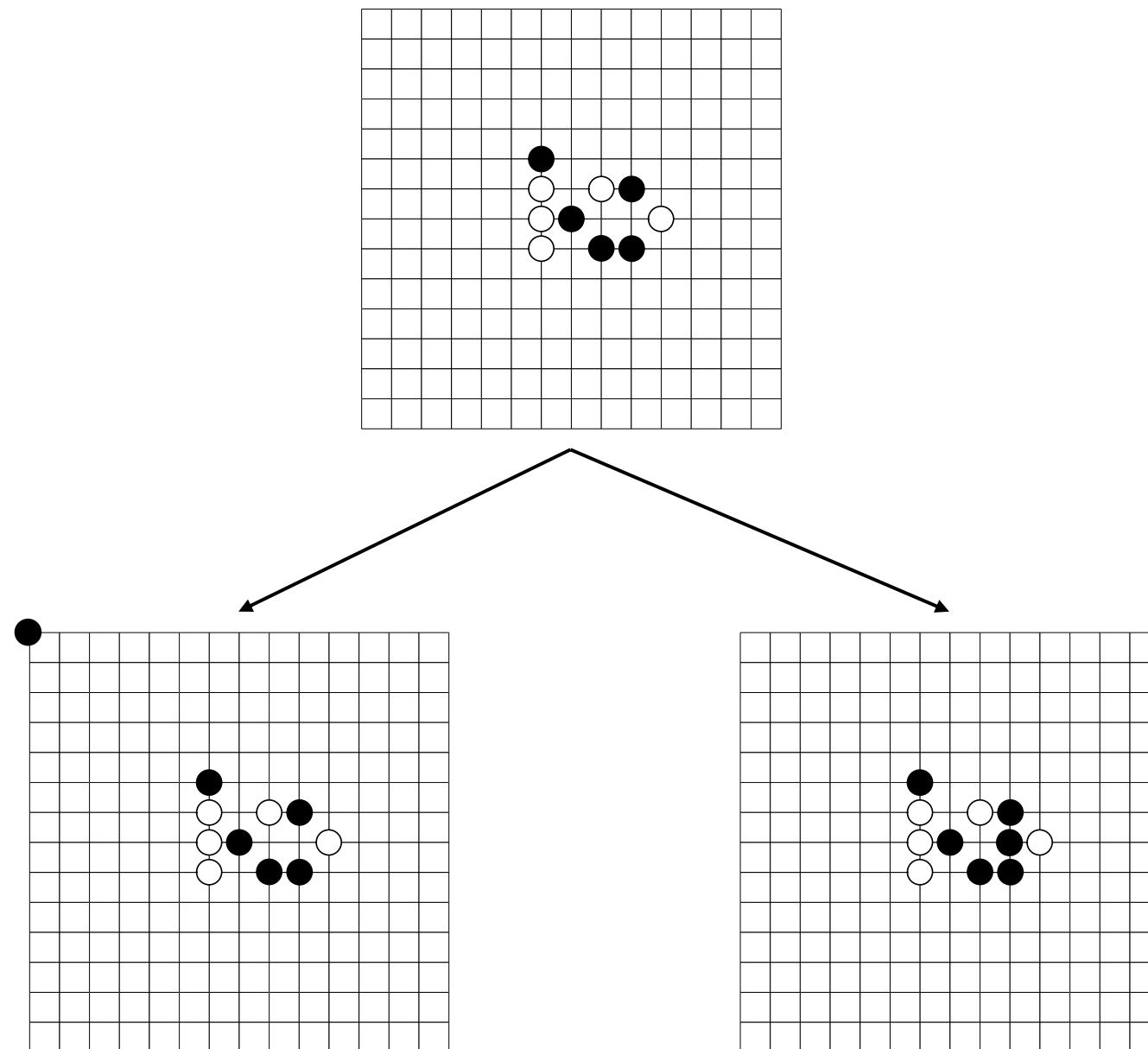
- Simulation, Expectation

- Steps

- Selection
 - Expansion
 - Simulation
 - Backpropagation



- Monte Carlo Tree Search
 - Simulation, Expectation
 - Steps
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- Monte Carlo Tree Search

- Simulation, Expectation

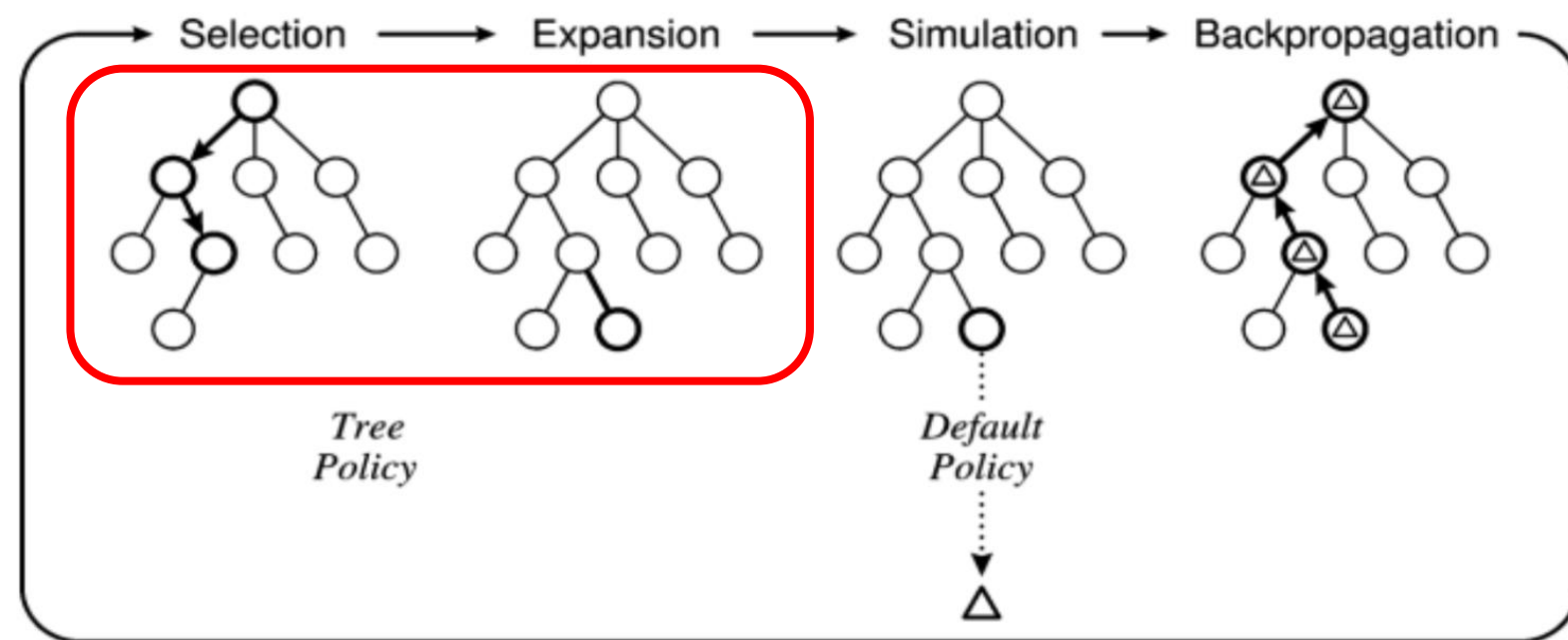
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- Monte Carlo Tree Search

- Simulation, Expectation

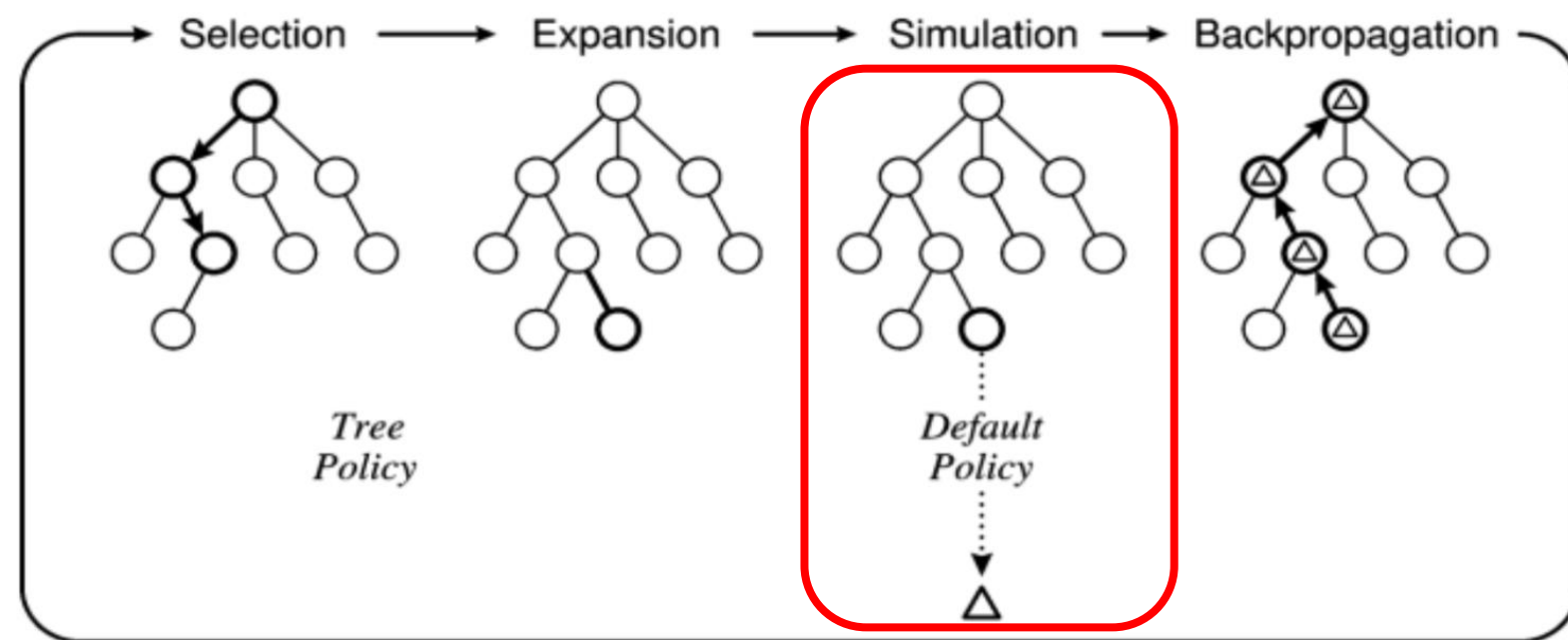
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- Monte Carlo Tree Search

- Simulation, Expectation

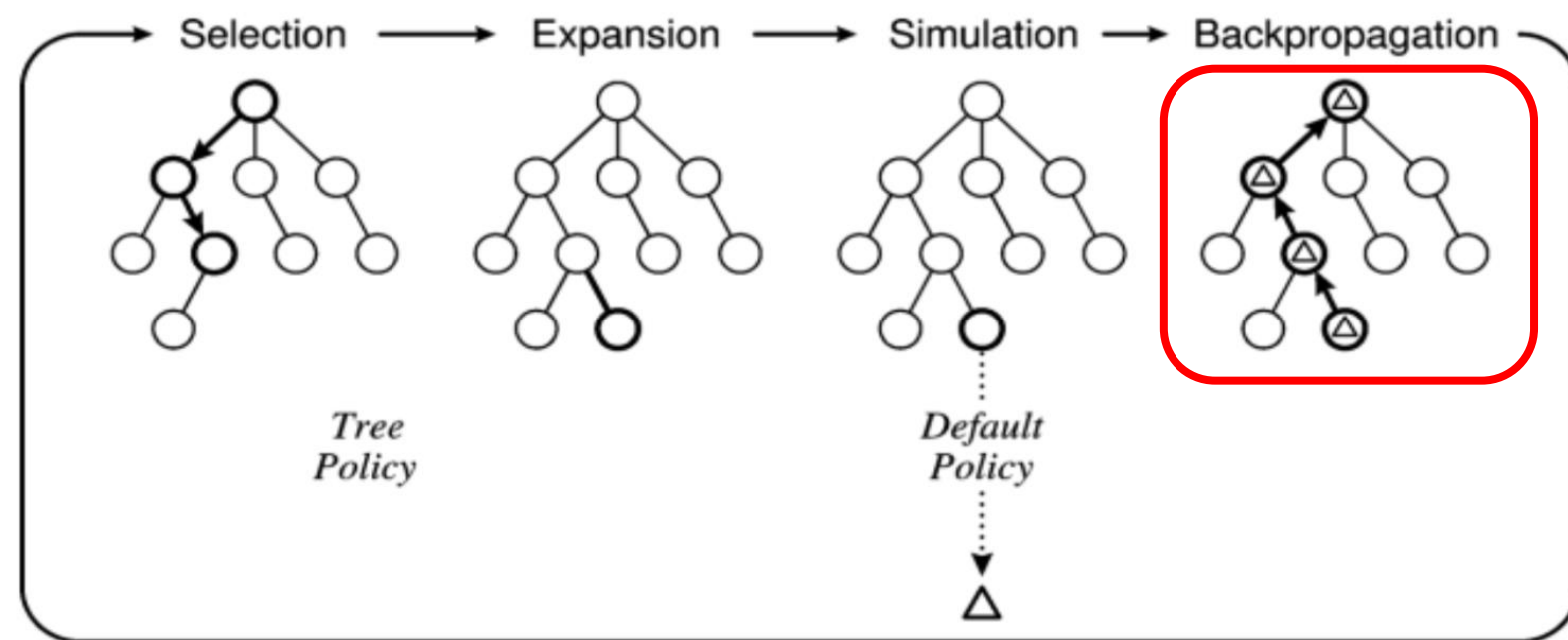
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- **Input** original state s_0
- **Output** action a corresponding to the highest value of MCTS

```
add Heuristic Knowledge;
obtain possible action moves  $M$  from state  $s_0$ ;
for each move  $m$  in moves  $M$  do
    reward  $r_{total} \leftarrow 0$ ;
    while simulation times < assigned times do
        reward  $r \leftarrow \text{Simulation}(s(m))$ ;
         $r_{total} \leftarrow r_{total} + r$ ;
        simulation times add one;
    end while
    add  $(m, r_{total})$  into  $data$ ;
end for each
return action  $\text{Best}(data)$ 
```

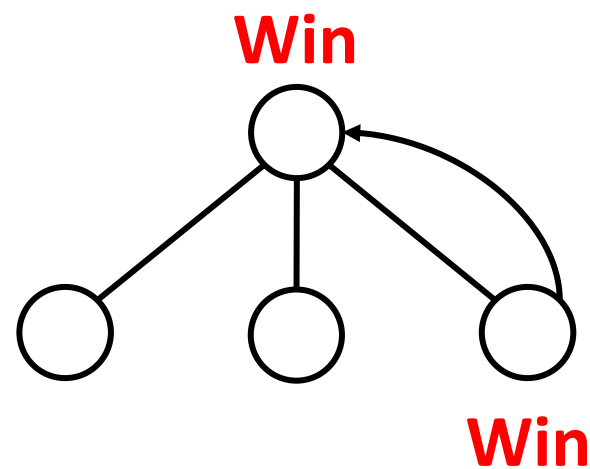
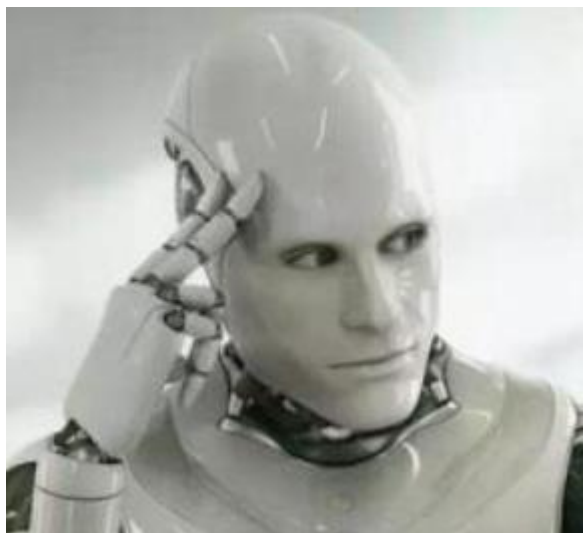
```
Simulation(state  $s_t$ )
    if ( $s_t$  is win and  $s_t$  is terminal) then return 1.0;
    else return 0.0;

    end if
    if ( $s_t$  satisfied with Heuristic Knowledge)
        then obtain forced action  $a_f$ ;
        new state  $s_{t+1} \leftarrow f(s_t, a_f)$ ;
    else choose random action  $a_r \in$  untried actions;
        new state  $s_{t+1} \leftarrow f(s_t, a_r)$ ;
    end if
    return Simulation( $s_{t+1}$ )
```

```
Best( $data$ )
    return action  $a$  //the maximum  $r_{total}$  of  $m$  from data
```

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- When will the Black win?

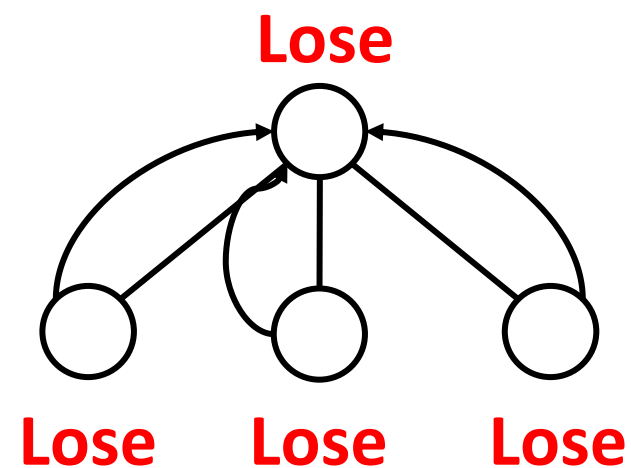


Louis Victor Allis.

Searching for Solutions in Games and Artificial Intelligence.

1994.

- When will the Black lose?



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- Board situation: Win, Lose, Unknown
- 2 Nodes:
 - Black Turn (OR)
 - Win if there is an action (White take) leading to Black win
 - Lose if all actions leading to Black lose
 - White (AND)
 - Win if all actions leading to Black win
 - Lose if there is an action leading to Black lose

- Board situation: Win, Lose, Unknown
- 2 Nodes: **Win or Lose: BLACK's View**
 - Black Turn (OR)
 - Win if there is an action (White take) leading to Black win
 - Lose if all actions leading to Black lose
 - White (AND)
 - Win if all actions leading to Black win
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Solve Gomoku: Proof-Number Search



- AND/OR Tree

- 3 Values: true, false, unknown

- 2 Nodes: AND, OR

- Black: OR

- Win if one child is win

- Unknown if no win and has unknown

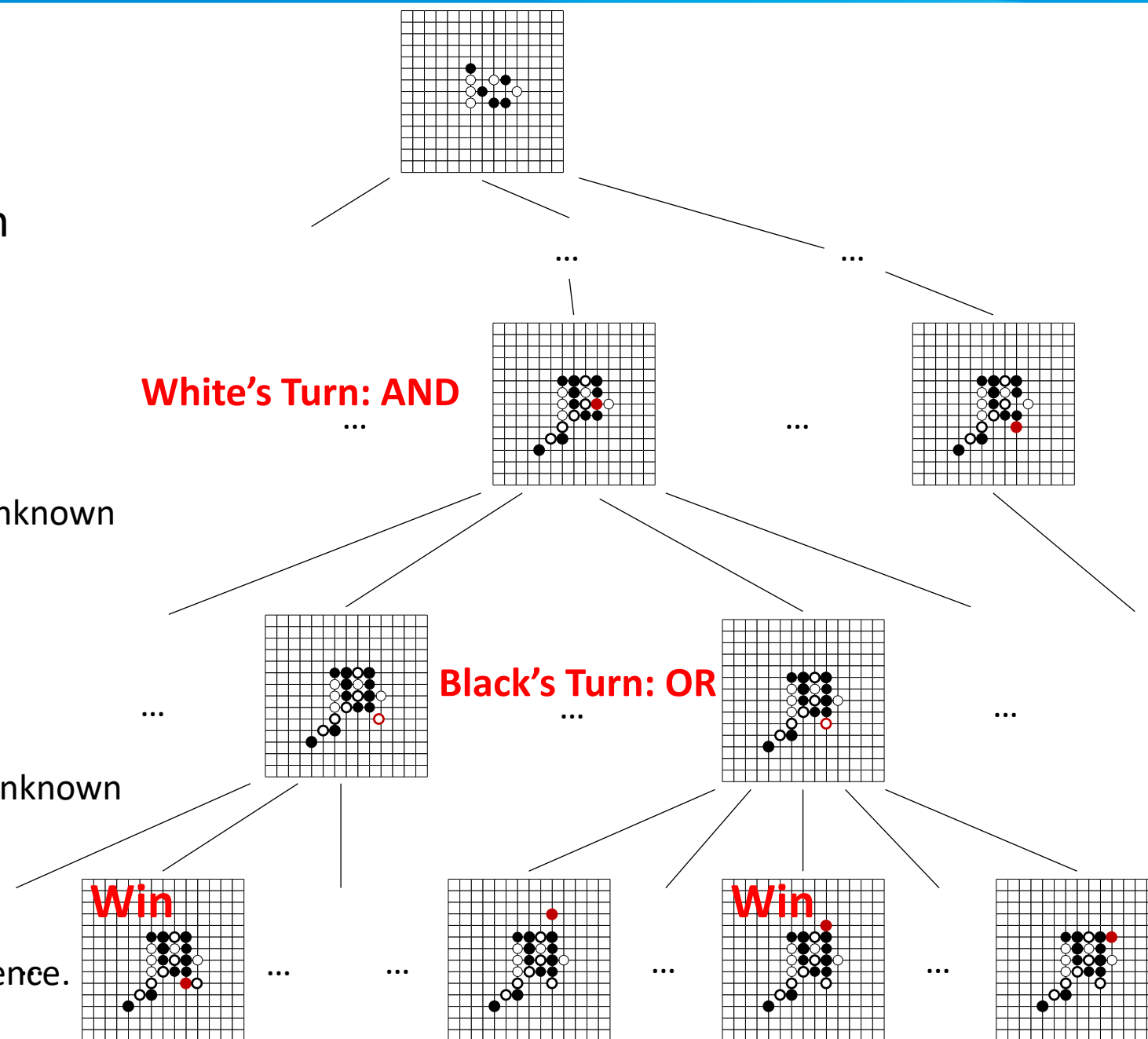
- Lose if all children are lose

- White: AND

- Lose if one child is lose

- Unknown if no lose and has unknown

- Win if all children are win



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- AND/OR Tree

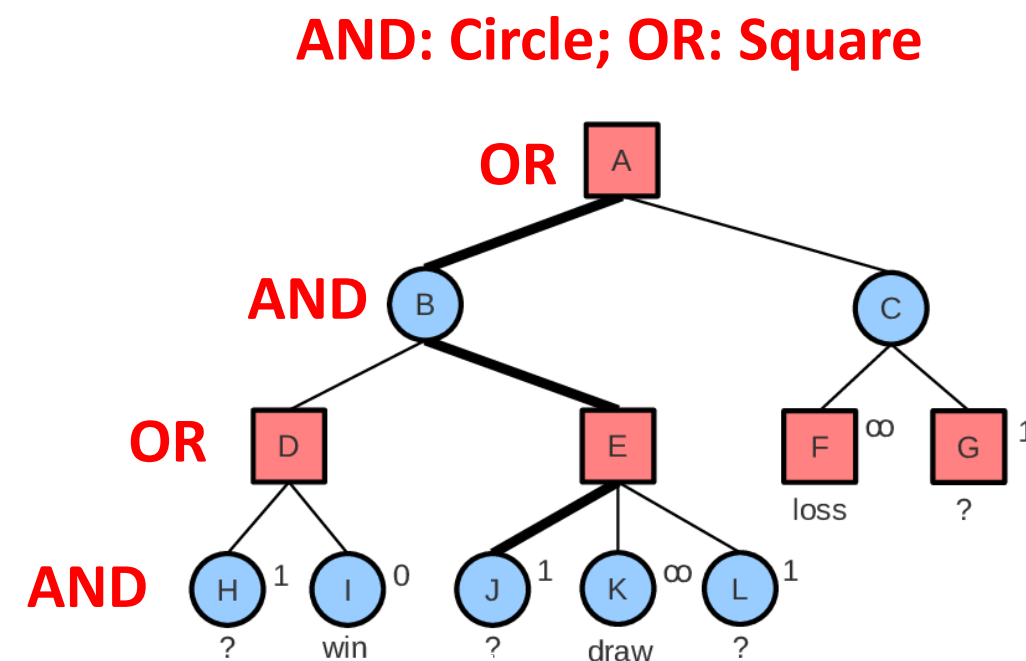
- Proof set

- Proof set: a set of frontier nodes S is a proof set if proving all nodes within S proves T
 - The proof number of T is defined as the cardinality of the smallest proof set of T

- Disproof set

- State of leaf nodes

- Win: 0, ∞
 - Lose: ∞ , 0
 - Unknown: 1, 1



- AND/OR Tree

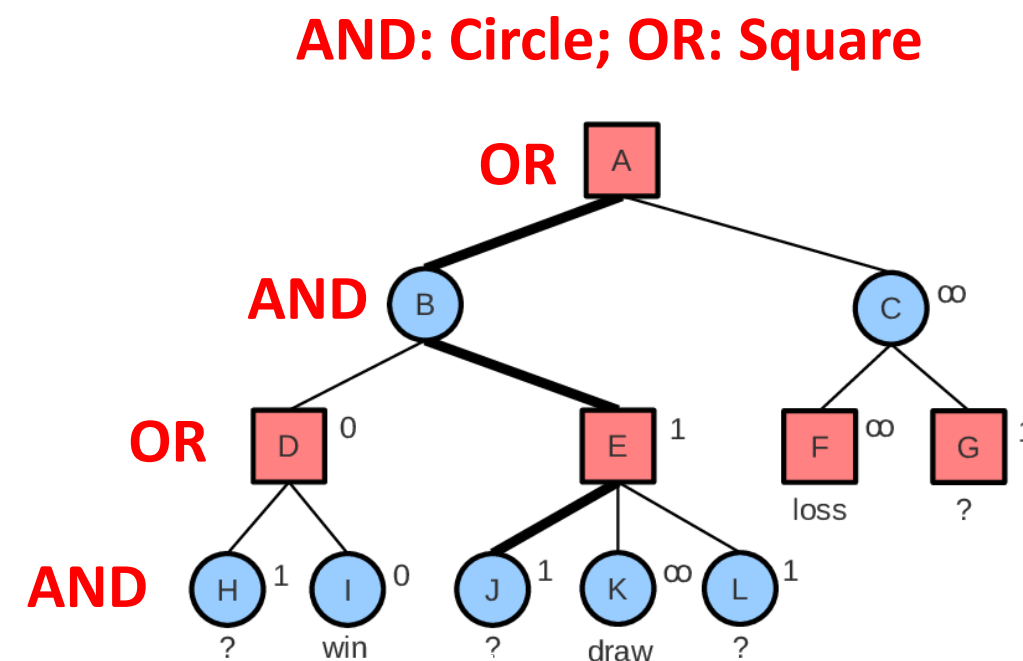
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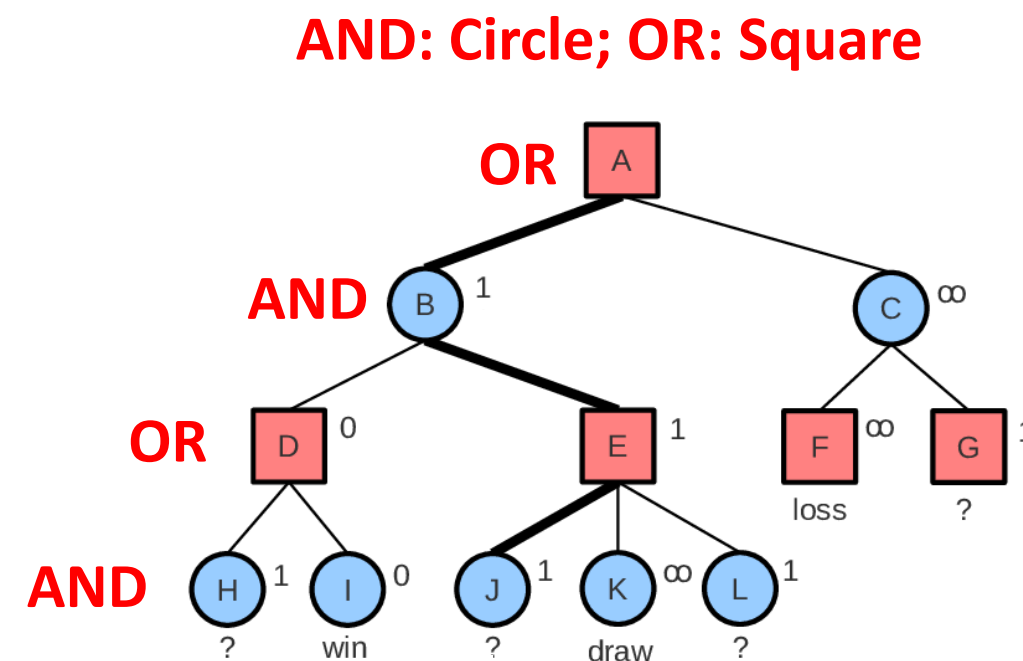
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- AND/OR Tree

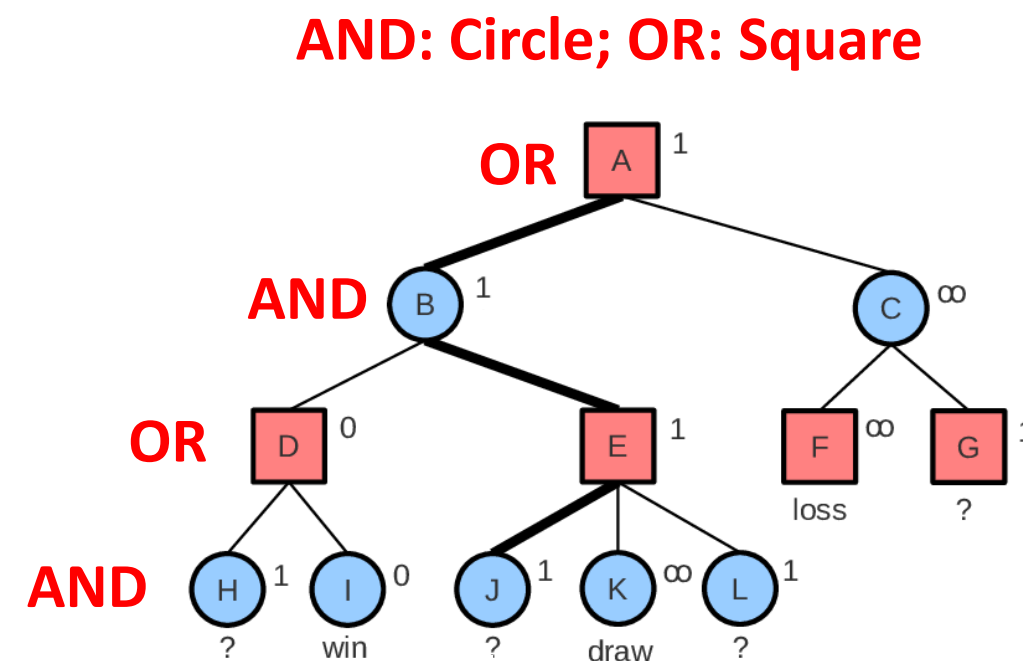
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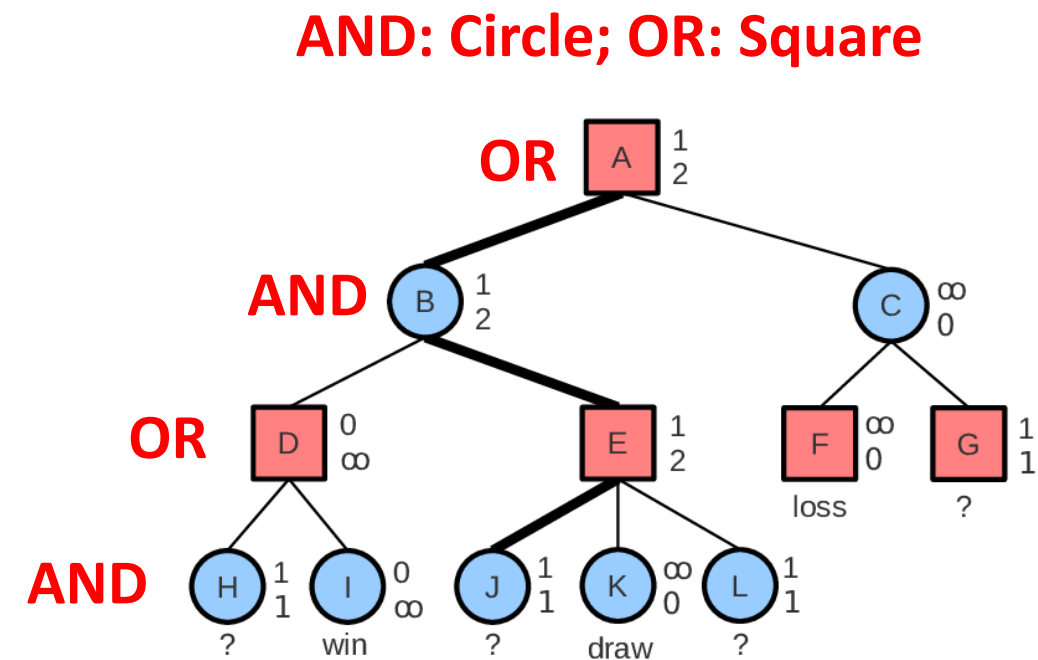
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- State of leaf nodes

- Win: 0, ∞
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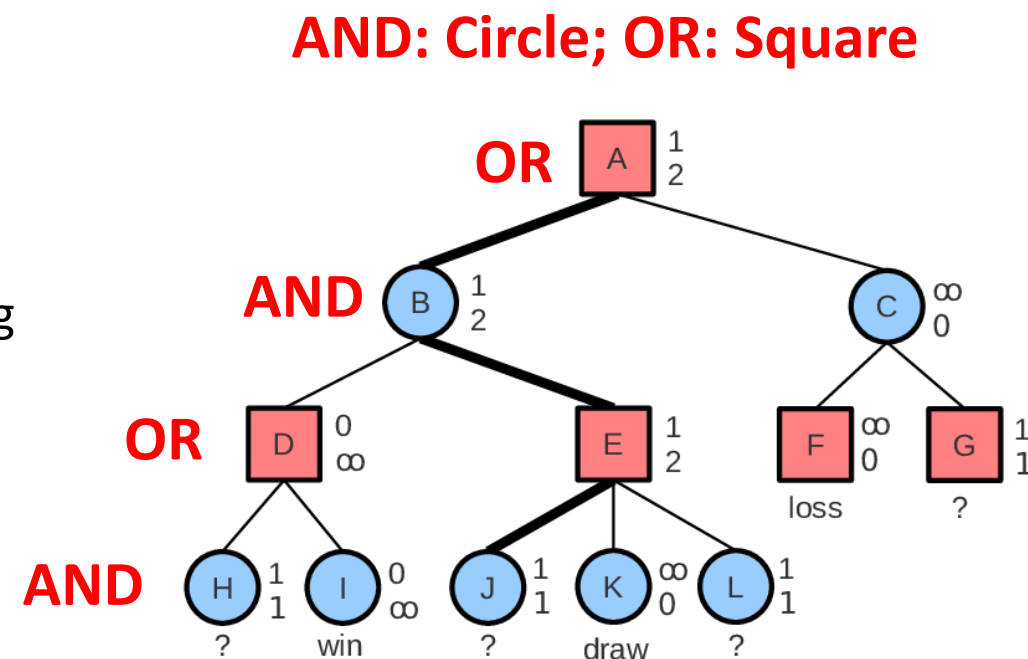


- AND/OR Tree
 - Set proof number
 - AND: sum OR: min
 - Set disproof number
 - AND: min OR: sum



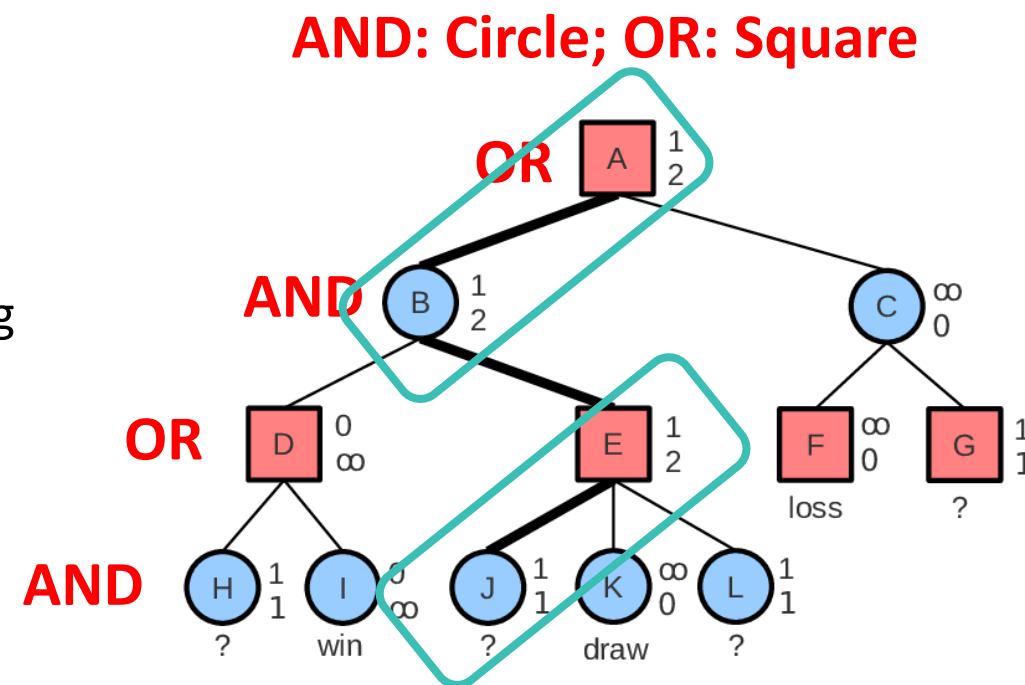
■ AND/OR Tree

- Most-proving node
 - Proved: each pair consisting of a smallest proof set and a smallest disproof set has a non-empty intersection.
 - i.e. There must exist at least one most-proving node.

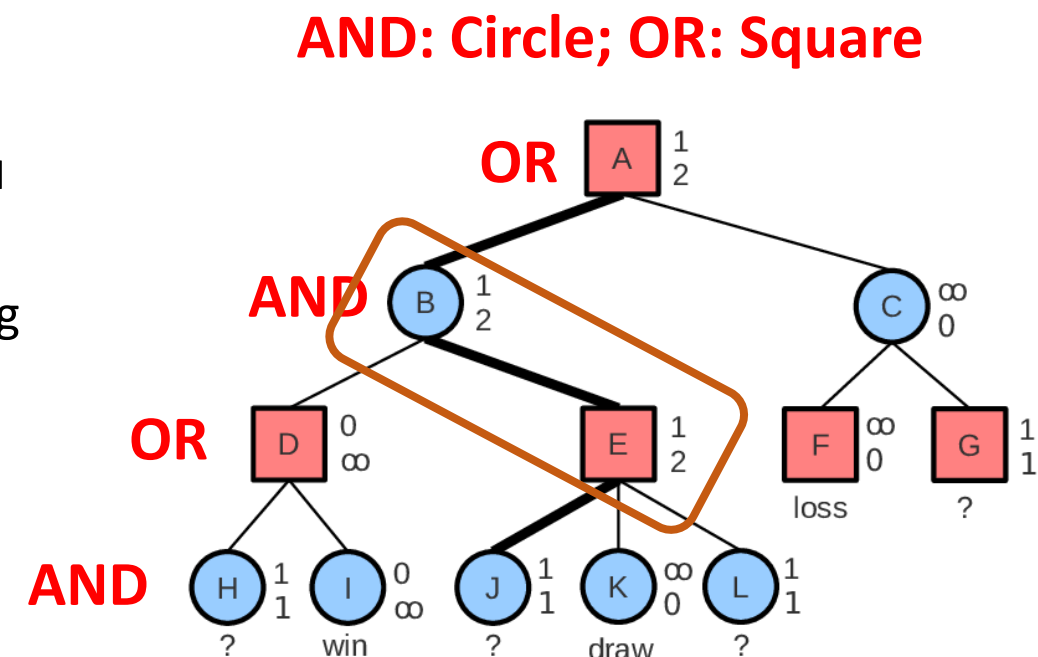


■ AND/OR Tree

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- AND/OR Tree
 - Most-proving node
 - Proved: each pair consisting of a smallest proof set and a smallest disproof set has a non-empty intersection.
 - i.e. There must exist at least one most-proving node.



■ Algorithm

```
procedure ProofNumberSearch(root);  
  Evaluate(root);  
  SetProofAndDisproofNumbers(root);  
  while root.proof  $\neq$  0 and root.disproof  $\neq$  0 and  
    ResourcesAvailable() do  
    mostProvingNode := SelectMostProving(root);  
    DevelopNode(mostProvingNode);  
    UpdateAncestors(mostProvingNode)  
  od ;  
  if root.proof = 0 then root.value := true  
  elseif root.disproof = 0 then root.value := false  
  else root.value := unknown  
  fi  
end
```

```
function SelectMostProving(node);  
  while node.expanded do  
    case node.type of  
      or :  
        i := 1;  
        while node.children[i].proof  $\neq$  node.proof do  
          i := i+1  
        od  
      and :  
        i := 1;  
        while node.children[i].disproof  $\neq$  node.disproof do  
          i := i+1  
        od  
      esac ;  
      node := node.children[i]  
    od ;  
  return node  
end
```

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■ Algorithm

```
procedure SetProofAndDisproofNumbers(node);  
  if node.expanded then  
    case node.type of  
      and :  
        node.proof :=  $\Sigma_{N \in \text{Children}(\text{node})} N.\text{proof}$ ;  
        node.disproof :=  $\text{Min}_{N \in \text{Children}(\text{node})} N.\text{disproof}$   
      or :  
        node.proof :=  $\text{Min}_{N \in \text{Children}(\text{node})} N.\text{proof}$ ;  
        node.disproof :=  $\Sigma_{N \in \text{Children}(\text{node})} N.\text{disproof}$   
    esac  
  elseif node.evaluated then  
    case node.value of  
      false : node.proof :=  $\infty$ ; node.disproof := 0  
      true : node.proof := 0; node.disproof :=  $\infty$   
      unknown : node.proof := 1; node.disproof := 1  
    esac  
  else node.proof := 1; node.disproof := 1  
  fi  
end
```

```
procedure DevelopNode(node);  
  GenerateAllChildren(node);  
  for i := 1 to node.numberOfChildren do  
    Evaluate(node.children[i]);  
    SetProofAndDisproofNumbers(node.children[i])  
  od  
end
```

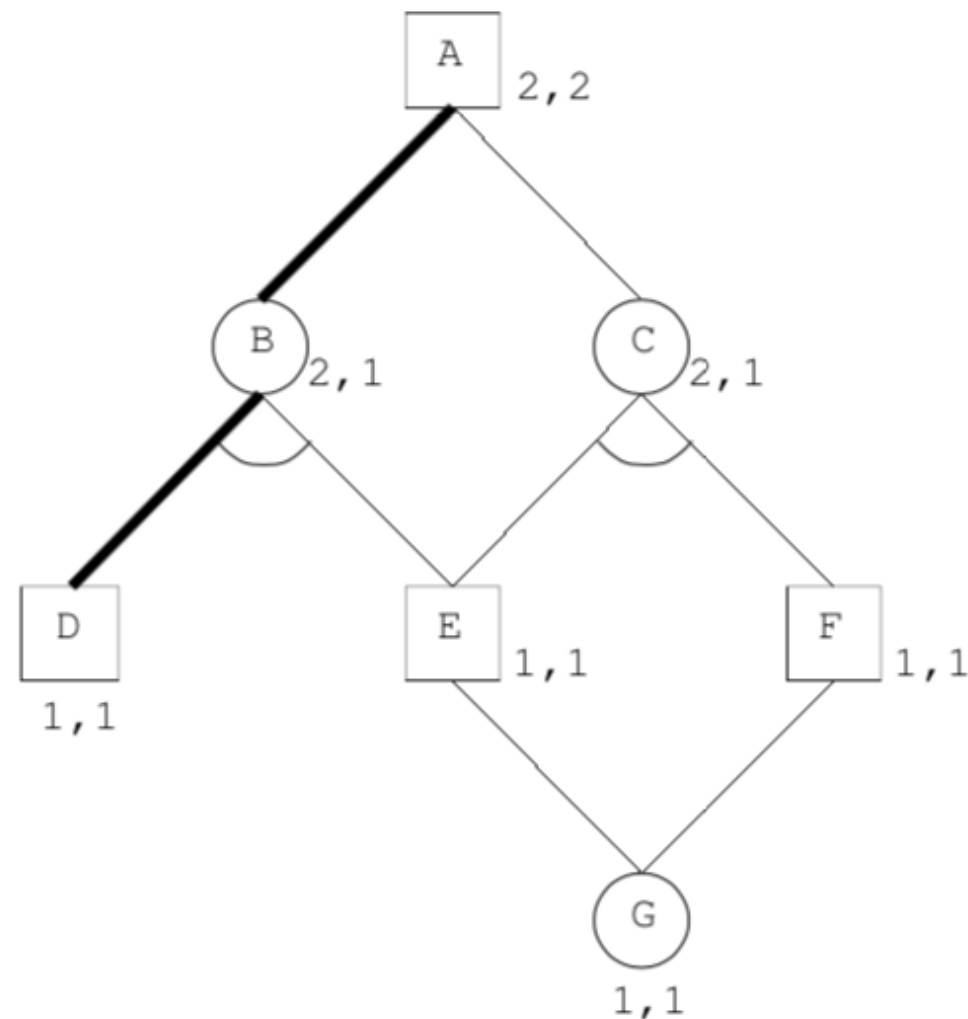
```
procedure UpdateAncestors(node);  
  while node  $\neq$  nil do  
    SetProofAndDisproofNumbers(node);  
    node := node.parent  
  od  
end
```

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- Transposition
 - Hash table
 - Directed Acyclic Graphs



Solve Gomoku: Monte Carlo Tree Search



- Monte Carlo Tree Search

- Simulation, Expectation

- Steps

- Selection

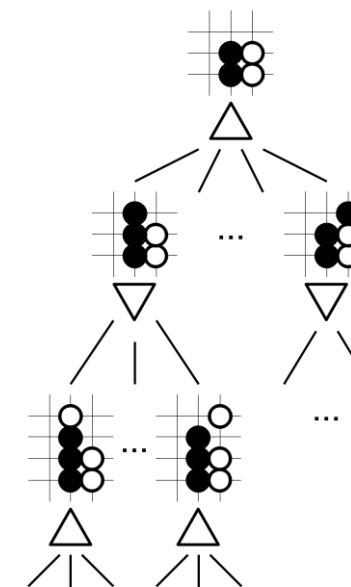
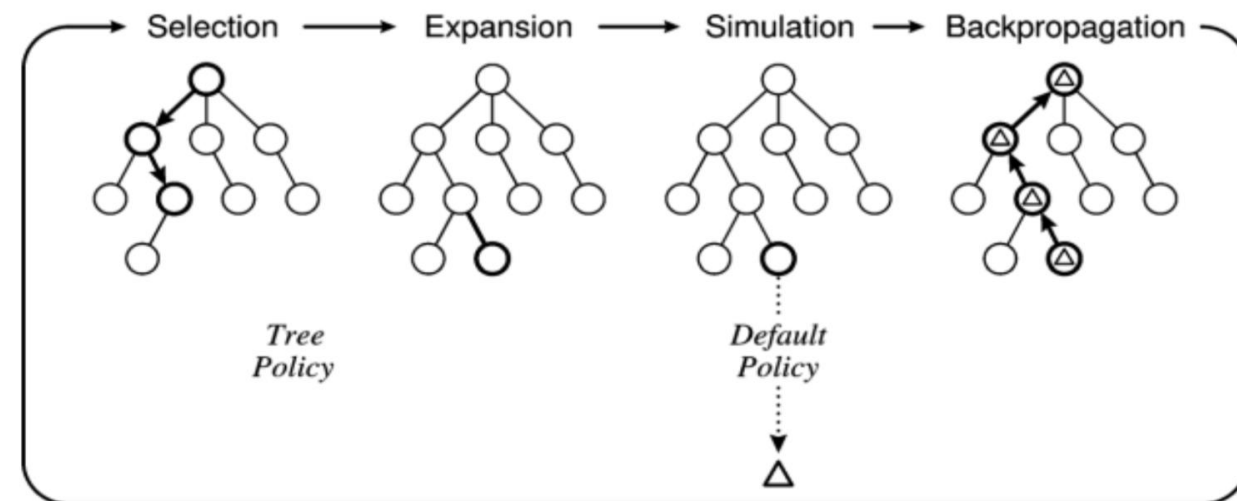
Proof-Number Search

- Expansion

Threat-Space Search, Genetic Algorithm

- Simulation

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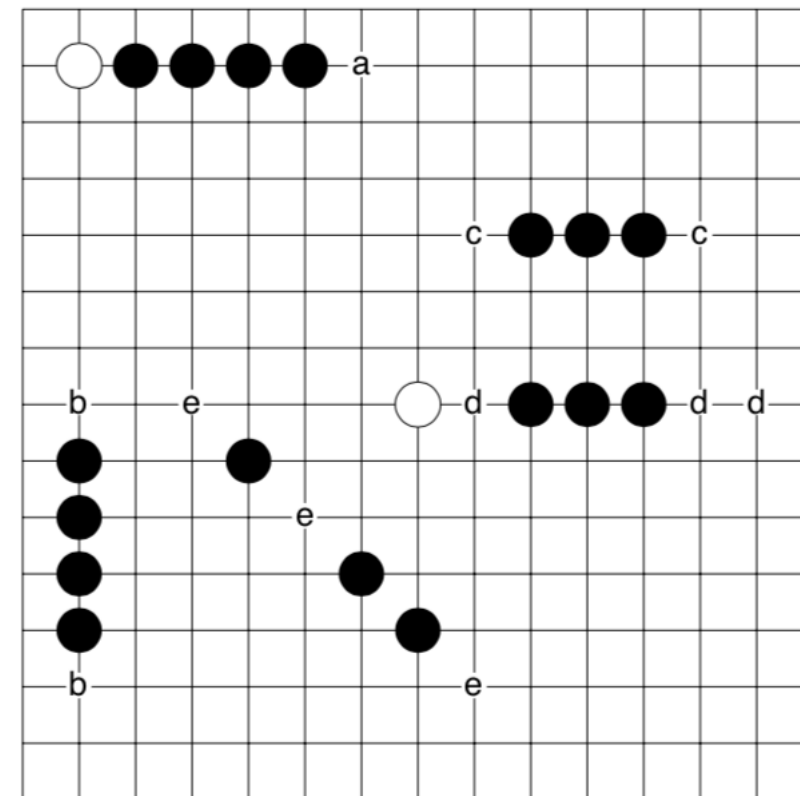
Solve Gomoku: Threat-Space Search



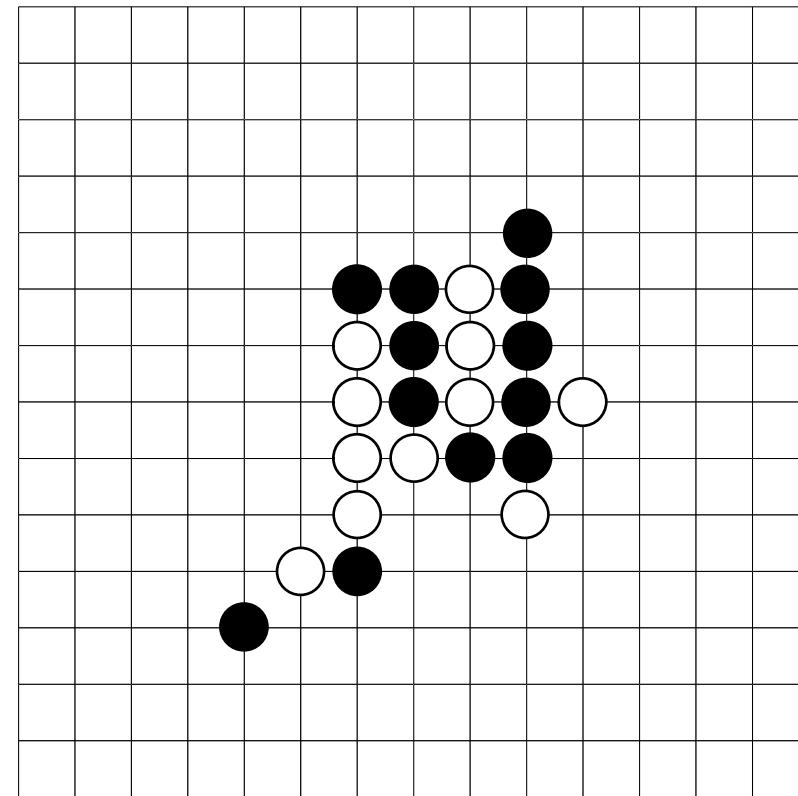
- Goal: Five in a row

Four in a row
Three in a row
... } Threat

Threat Sequence



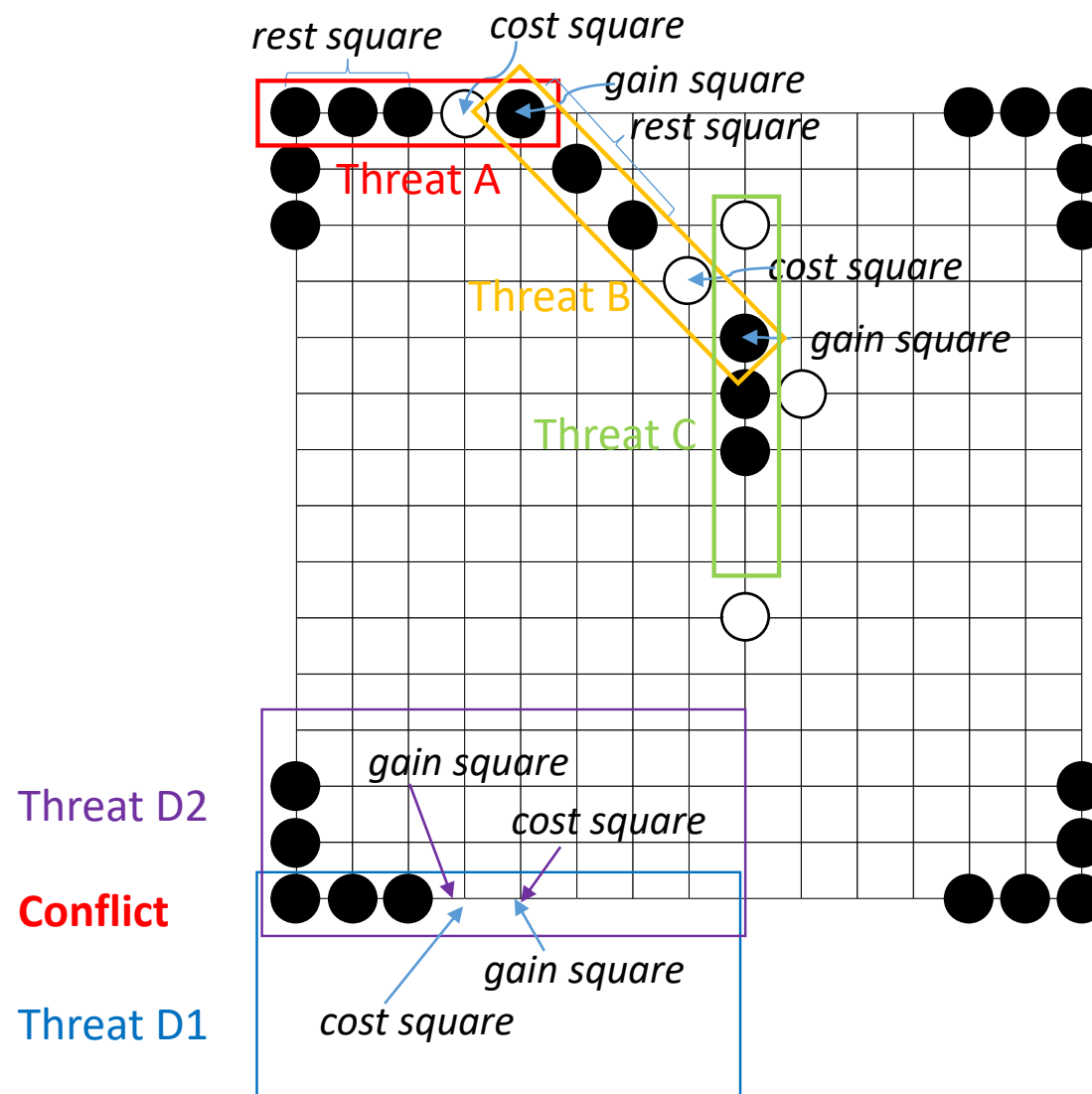
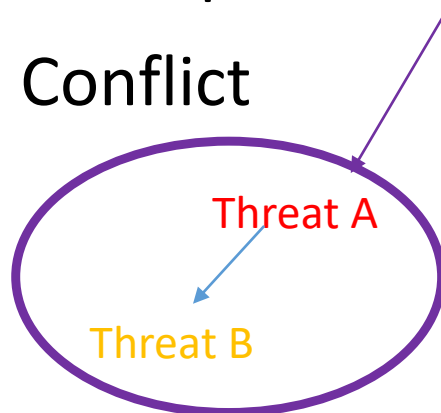
- Threat Sequence
- Winning Threat Sequence



Solve Gomoku: Threat-Space Search



- Gain square
- Cost square
- Rest square
- Dependent
 - Dependency tree
- Conflict



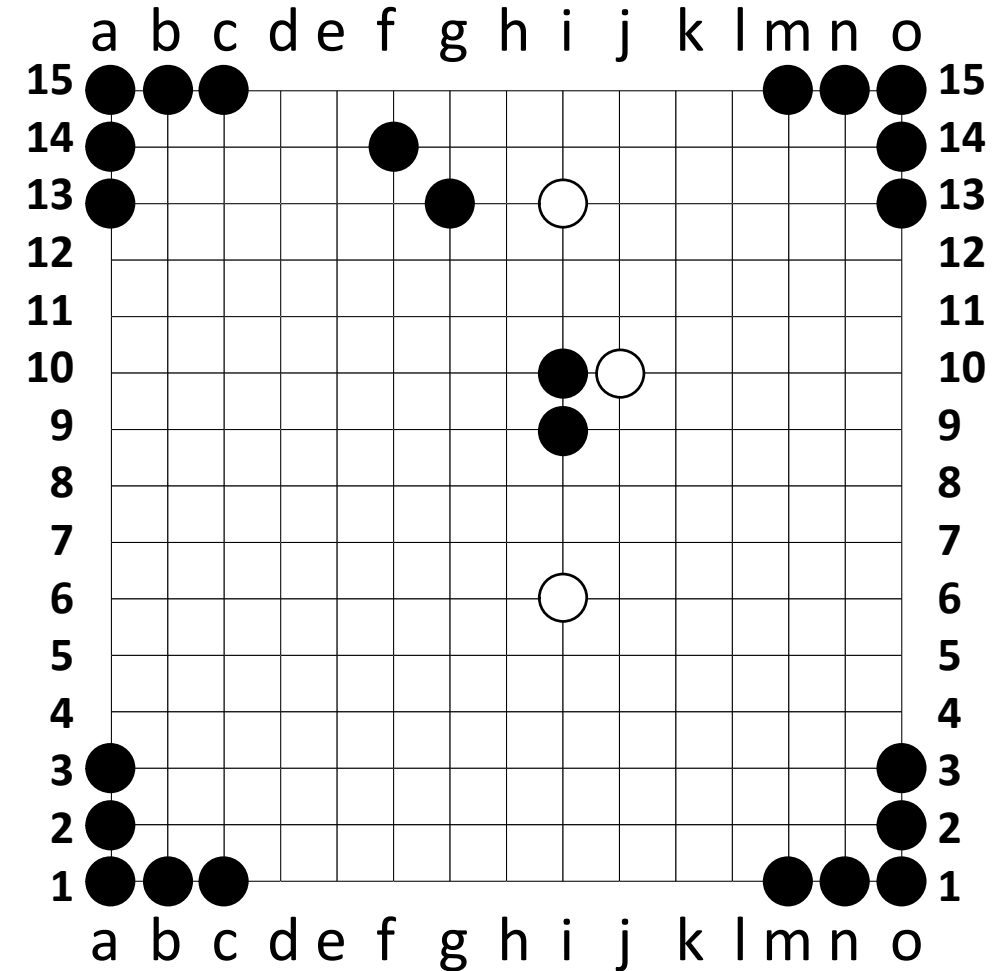
Solve Gomoku: Threat-Space Search



■ Threat Search tree:

- Threat A being independent of threat B is not allowed to occur in the search tree of threat B.
- Only threats for the attacker are included.

Depth	Type of threat	Gain square	Cost squares
1	Four	<i>l15</i>	<i>k15</i>
1	Four	<i>k15</i>	<i>l15</i>
1	Four	<i>e15</i>	<i>d15</i>
2	Four	<i>i11</i>	<i>h12</i>
3	Straight Four	<i>i8</i>	<i>i7</i>
2	Four	<i>h12</i>	<i>i11</i>
1	Four	<i>d15</i>	<i>e15</i>
1	Four	<i>o12</i>	<i>o11</i>
1	Four	<i>o11</i>	<i>o12</i>
1	Four	<i>a12</i>	<i>a11</i>
1	Four	<i>a11</i>	<i>a12</i>
1	Three	<i>i11</i>	<i>i7,i8,i12</i>
2	Four	<i>h12</i>	<i>e15</i>
2	Four	<i>e15</i>	<i>h12</i>
3	Five	<i>d15</i>	
1	Three	<i>i8</i>	<i>i7,i11,i12</i>
1	Four	<i>o5</i>	<i>o4</i>
1	Four	<i>o4</i>	<i>o5</i>
1	Four	<i>l1</i>	<i>k1</i>
1	Four	<i>k1</i>	<i>l1</i>
1	Four	<i>e1</i>	<i>d1</i>
1	Four	<i>d1</i>	<i>e1</i>
1	Four	<i>a5</i>	<i>a4</i>
1	Four	<i>a4</i>	<i>a5</i>



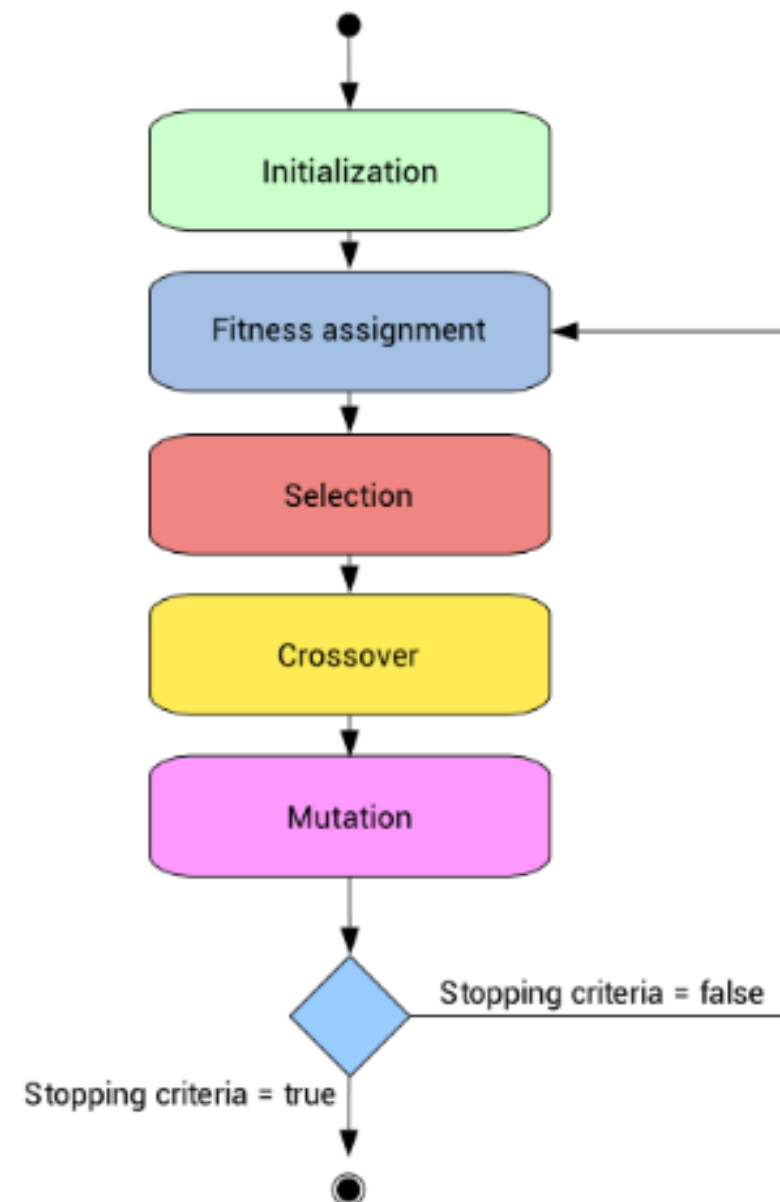
- Victoria
 - Threat-space search
 - Proof-number search
- Threat-Space Search
 - a module capable of quickly determining whether a winning threat sequence exists
 - used as a first evaluation function
 - Win for the attacker
 - No win: proof-number search
 - a heuristic evaluation procedure

- Gomoku Rule
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Solve Gomoku: Genetic Algorithm



- Initialization: Coding Scheme
- Fitness assignment
- Selection
- Crossover
- Mutation



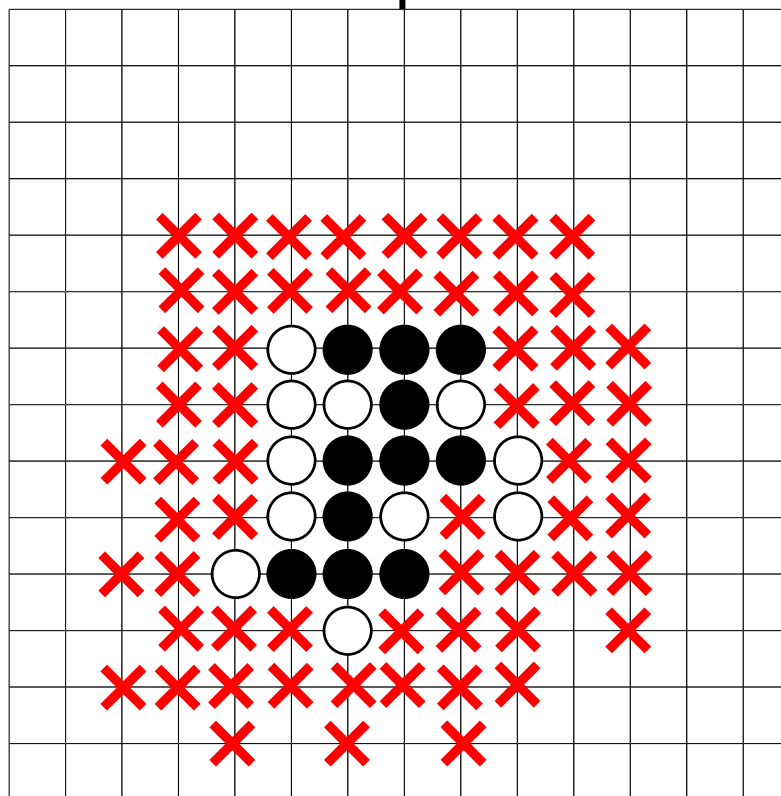
- Initialization: Coding Scheme
 - Coordinates of consecutive K steps

- N sequences

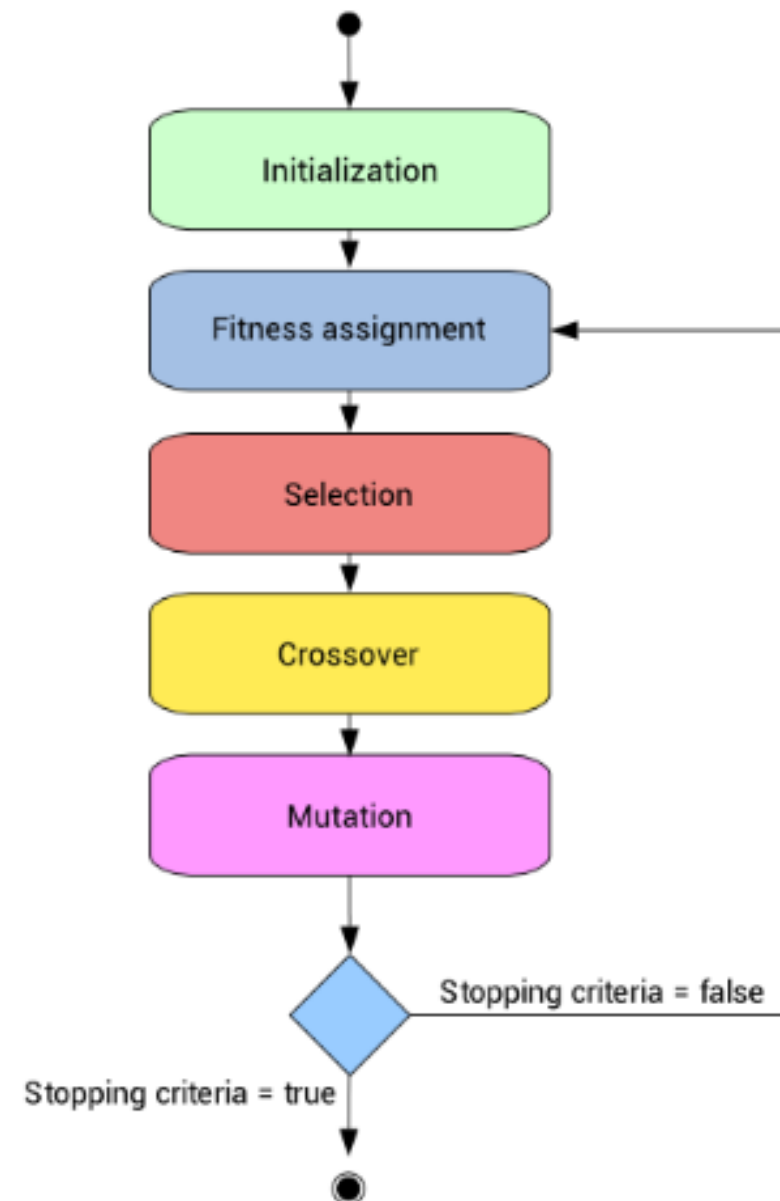
- Representation:

- $A_2A_1A_7A_3A_8A_5A_6$
- $A_5A_3A_1A_4A_9A_2A_8$
- $A_1A_6A_8A_2A_5A_3A_4$
- $A_2A_9A_3A_4A_7A_8A_6$
- $A_1A_4A_7A_6A_5A_8A_2$

Both you and the enemy



- Initialization: Coding Scheme
- Fitness assignment
 - Example:
$$f(s) = 4800 * (\text{number of four structures in neighborhood}) + 97 * (\text{number of three structures in neighborhood}) + 17 * (\text{number of two structures in neighborhood})$$
- Selection



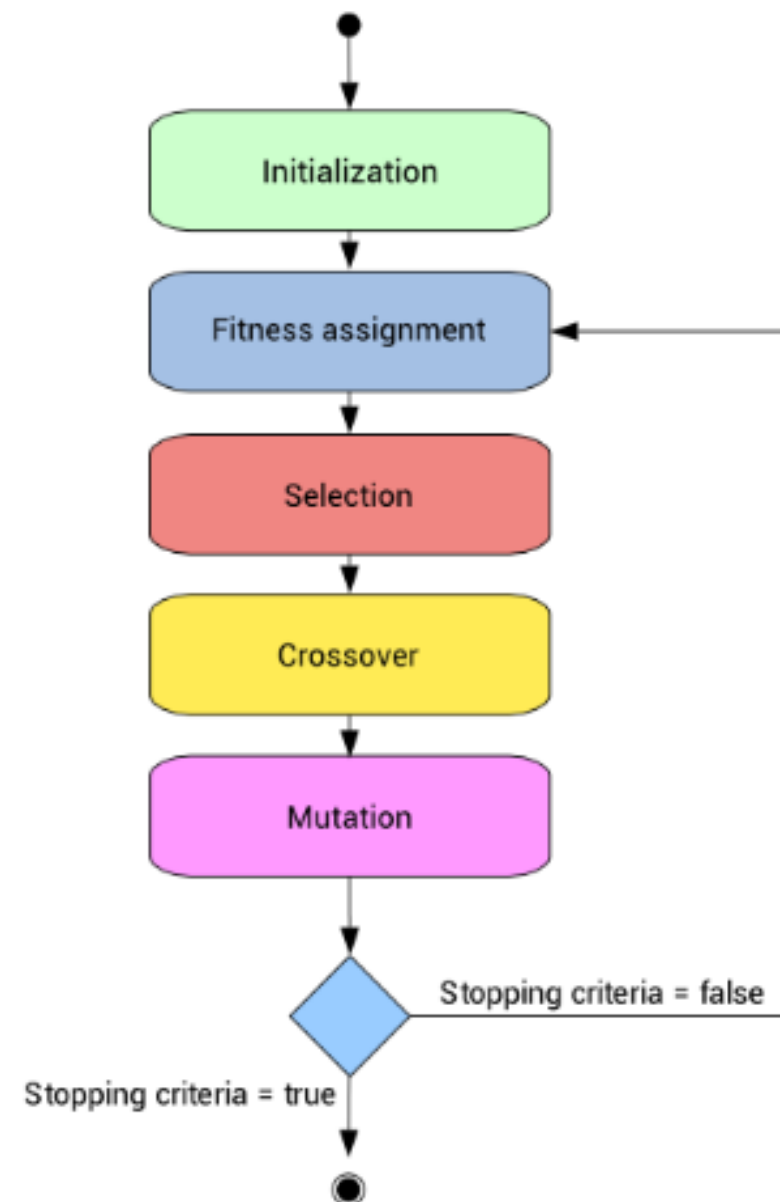
- Initialization: Coding Scheme
- Fitness assignment
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- Crossover

■ Parents: $A_2A_1A_7A_3A_8A_5A_6$
 $A_5A_3A_1A_4A_9A_2A_8$

■ Children: $A_2A_1A_7A_3A_9A_2A_8$
 $A_5A_3A_1A_4A_8A_5A_6$

- Mutation

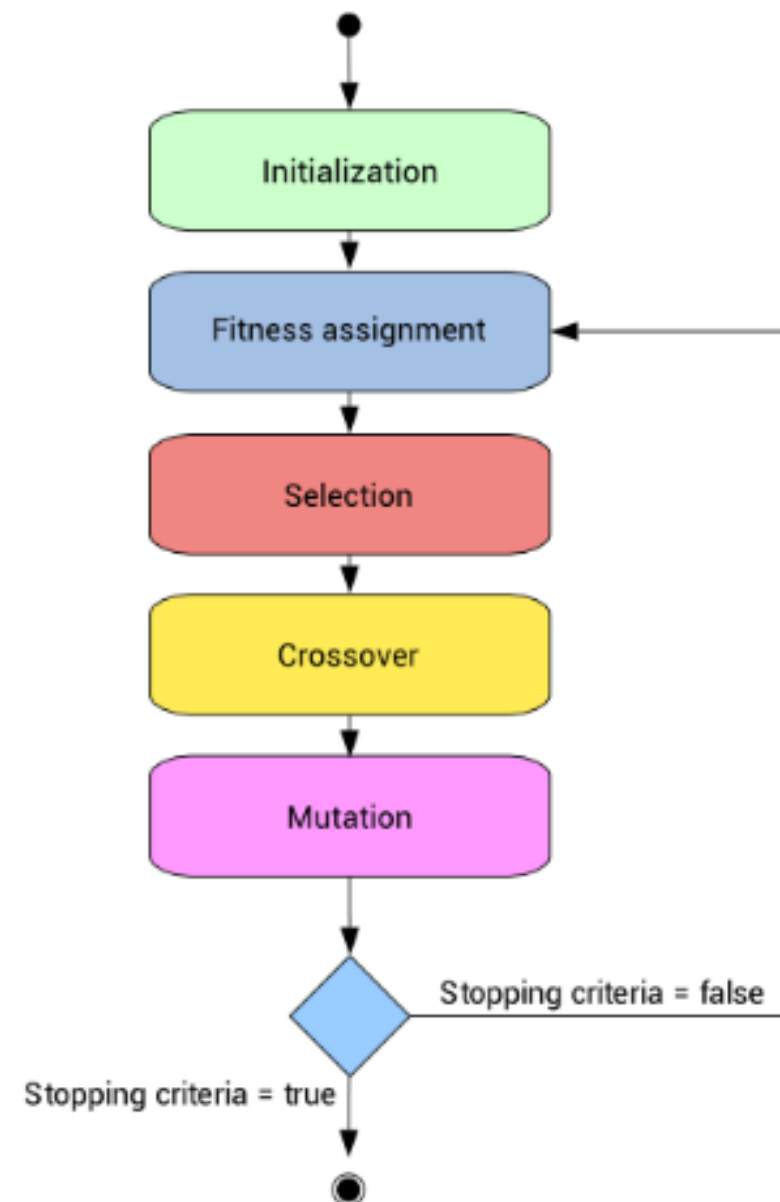
$A_2A_9A_3A_4A_7A_8A_6$
 $A_2A_9A_7A_8A_3A_4A_6$



Solve Gomoku: Genetic Algorithm



- Initialization: Coding Scheme
- Fitness assignment
- Selection
- Crossover
- Mutation



- Gomoku manager
 - <http://gomocup.org/download-gomocup-manager/>
- AI
 - <http://gomocup.org/download-gomoku-ai/>
- Python Template
 - <https://github.com/stranskyjan/pbrain-pyrandom>
- Gomocup
 - <http://gomocup.org/>