

## Search 2

Shivaram Kalyanakrishnan

Department of Computer Science and Engineering  
Indian Institute of Technology Bombay

February 2023

# This Lecture

- Informed search (a.k.a. heuristic search)
- Application in game-playing

# This Lecture

- Informed search (a.k.a. heuristic search)
- Application in game-playing

## Generic Search Template: Pseudocode

- Primary data element is a **Node**, which is a tuple of the form  
 $(state, pathFromStartState, pathCost)$ .

## Generic Search Template: Pseudocode

- Primary data element is a **Node**, which is a tuple of the form  
 $(state, pathFromStartState, pathCost)$ .
- At every stage of the search,
  - some states have been **explored**
  - some states remain **unexplored**, and
  - The **Frontier** is a set of nodes due for imminent expansion.

## Generic Search Template: Pseudocode

- Primary data element is a **Node**, which a tuple of the form

$(state, pathFromStartState, pathCost)$ .

- At every stage of the search,
  - some states have been **explored**
  - some states remain **unexplored**, and
  - The **Frontier** is a set of nodes due for imminent expansion.

```
Frontier  $\leftarrow \{Node(startState, (startState), 0)\}.$ 
```

```
Repeat for ever:
```

```
    Select a node  $n$  from Frontier.
```

```
    //Expand  $n$ .
```

```
    If  $isGoal(n.state)$ :
```

```
        Return  $n$ .
```

```
    For each action  $a$  available from  $n.state$ :
```

```
         $s \leftarrow NextState(n.state, a).$ 
```

```
         $c \leftarrow Cost(n.state, a).$ 
```

```
         $n' \leftarrow Node(s, n.path + (a, s), n.pathCost + c).$ 
```

```
        Merge  $n'$  with Frontier. //Typically insertion; might allow deletions.
```

## Generic Search Template: Pseudocode

- Primary data element is a **Node**, which a tuple of the form

$(state, pathFromStartState, pathCost)$ .

- At every stage of the search,
  - some states have been **explored**
  - some states remain **unexplored**, and
  - The **Frontier** is a set of nodes due for imminent expansion.

```
Frontier  $\leftarrow \{Node(startState, (startState), 0)\}.$ 
```

```
Repeat for ever:
```

```
    Select a node  $n$  from Frontier. //How is this selection made?
```

```
    //Expand  $n$ .
```

```
    If  $isGoal(n.state)$ :
```

```
        Return  $n$ .
```

```
    For each action  $a$  available from  $n.state$ :
```

```
         $s \leftarrow NextState(n.state, a).$ 
```

```
         $c \leftarrow Cost(n.state, a).$ 
```

```
         $n' \leftarrow Node(s, n.path + (a, s), n.pathCost + c).$ 
```

```
        Merge  $n'$  with Frontier. //Typically insertion; might allow deletions.
```

# Incorporating Domain Knowledge into Search

- Have to travel from Powai to Mahim.

Powai

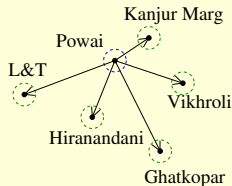


- Mahim



## Incorporating Domain Knowledge into Search

- Have to travel from Powai to Mahim.

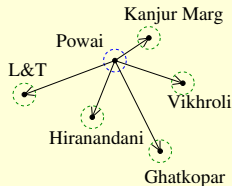


- Mahim

- First you expand the Powai node.

## Incorporating Domain Knowledge into Search

- Have to travel from Powai to Mahim.

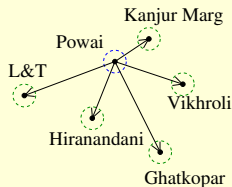


- Mahim

- First you expand the Powai node.
- Which node will **you** expand next?

## Incorporating Domain Knowledge into Search

- Have to travel from Powai to Mahim.



- Mahim

- First you expand the Powai node.
- Which node will **you** expand next?
- L&T and Hiranandani are **geographically** closer to Mahim: should that count?

## Heuristic Functions and A<sup>\*</sup> Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .

## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

- Recall that in LCFS, we expand

$$\operatorname{argmin}_{n \in \text{Frontier}} g(n).$$

## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

- Recall that in LCFS, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} g(n).$$

- In A\* search, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} (g(n) + h(n)).$$

## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

- Recall that in LCFS, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} g(n).$$

- In A\* search, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} (g(n) + h(n)).$$

- $g(n)$  summarises the past (known);  $h(n)$  anticipates the future (unknown).



## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

- Recall that in LCFS, we expand

$$\operatorname{argmin}_{n \in \text{Frontier}} g(n).$$

- In A\* search, we expand

$$\operatorname{argmin}_{n \in \text{Frontier}} (g(n) + h(n)).$$

- $g(n)$  summarises the past (known);  $h(n)$  anticipates the future (unknown).
- The addition of  $h(n)$  makes A\* an **informed** or **heuristic** search algorithm.

## Heuristic Functions and A\* Search Algorithm

- A **heuristic** function  $h(n)$  is a guess of  $c^*(n)$ , the optimal path-cost-to-goal of (the state in) node  $n$ .
- $h(n)$  is usually easy to compute. On the previous slide, we implicitly used straight line distance:

$$h(n) = \sqrt{(n.state.x - Mahim.x)^2 + (n.state.y - Mahim.y)^2}.$$

- Recall that in LCFS, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} g(n).$$

- In A\* search, we expand

$$\underset{n \in \text{Frontier}}{\operatorname{argmin}} (g(n) + h(n)).$$

- $g(n)$  summarises the past (known);  $h(n)$  anticipates the future (unknown).
- The addition of  $h(n)$  makes A\* an **informed** or **heuristic** search algorithm.
- A\* search was originally conceived for robotic path planning.

# Admissible Heuristics

- A heuristic  $h$  is **admissible** if for all nodes  $n$ ,

$$0 \leq h(n) \leq c^*(n),$$

where  $c^*(n)$  is the optimal cost-to-goal of  $n.state$ .

# Admissible Heuristics

- A heuristic  $h$  is **admissible** if for all nodes  $n$ ,

$$0 \leq h(n) \leq c^*(n),$$

where  $c^*(n)$  is the optimal cost-to-goal of  $n.state$ .

- **Key result.** If  $A^*$  search is run using an admissible heuristic (and some minor technical conditions hold), then the **first goal node** it expands will have **optimal** path-cost from the start state (and the algorithm can terminate).

# Admissible Heuristics

- A heuristic  $h$  is **admissible** if for all nodes  $n$ ,

$$0 \leq h(n) \leq c^*(n),$$

where  $c^*(n)$  is the optimal cost-to-goal of  $n.state$ .

- **Key result.** If  $A^*$  search is run using an admissible heuristic (and some minor technical conditions hold), then the **first goal node** it expands will have **optimal** path-cost from the start state (and the algorithm can terminate).
- Is straight line distance an admissible heuristic for navigation?

# Admissible Heuristics

- A heuristic  $h$  is **admissible** if for all nodes  $n$ ,

$$0 \leq h(n) \leq c^*(n),$$

where  $c^*(n)$  is the optimal cost-to-goal of  $n.state$ .

- **Key result.** If  $A^*$  search is run using an admissible heuristic (and some minor technical conditions hold), then the **first goal node** it expands will have **optimal** path-cost from the start state (and the algorithm can terminate).
- Is straight line distance an admissible heuristic for navigation?  
Yes.

# Admissible Heuristics

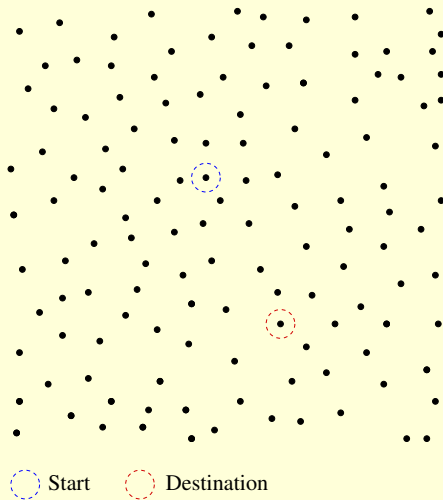
- A heuristic  $h$  is **admissible** if for all nodes  $n$ ,

$$0 \leq h(n) \leq c^*(n),$$

where  $c^*(n)$  is the optimal cost-to-goal of  $n.state$ .

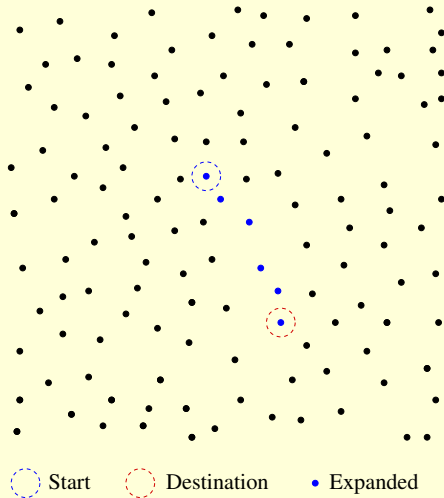
- **Key result.** If  $A^*$  search is run using an admissible heuristic (and some minor technical conditions hold), then the **first goal node** it expands will have **optimal** path-cost from the start state (and the algorithm can terminate).
- Is straight line distance an admissible heuristic for navigation?  
Yes.
- For a given task, which is the best heuristic function to use?

## Effect of Heuristic



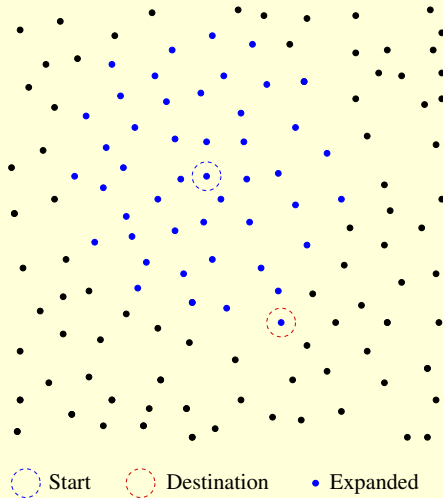


## Effect of Heuristic



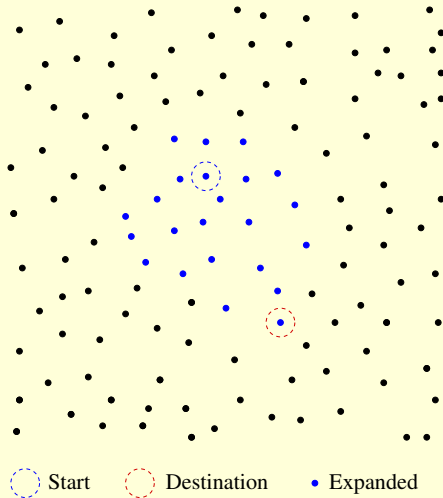
$h(n) = c^*(n)$ . Will only expand nodes along optimal path! But  $c^*(n)$  not known!

## Effect of Heuristic



$h(n) = 0$ . Identical to LCFS.

## Effect of Heuristic



Intermediate/typical  $h(n)$  expands fewer nodes than LCFS.

## Questions

- How to come up with an effective admissible heuristic for a task?

## Questions

- How to come up with an effective admissible heuristic for a task?

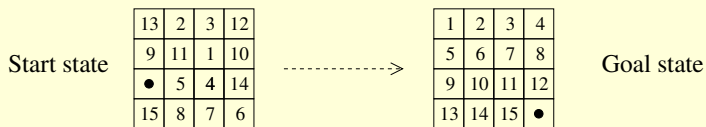
For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?

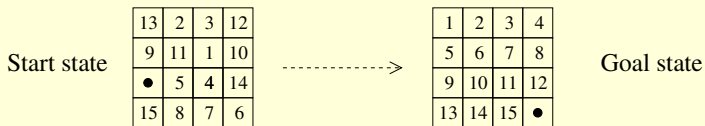


## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?



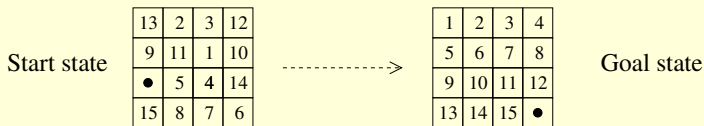
Sum of Manhattan distances between each number's position in start state and its position in goal state.

## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?



Sum of Manhattan distances between each number's position in start state and its position in goal state.

- Can we make do with inadmissible heuristics?

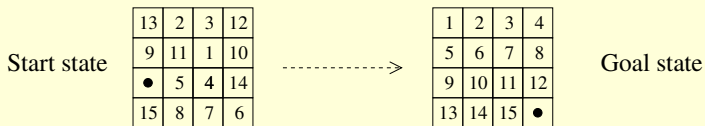


## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?



Sum of Manhattan distances between each number's position in start state and its position in goal state.

- Can we make do with inadmissible heuristics?

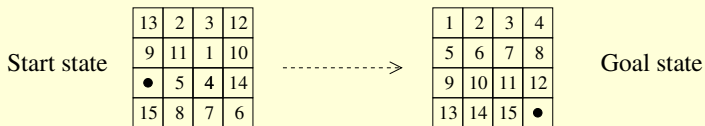
Yes—example coming up in next section. But try to avoid.

## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?



Sum of Manhattan distances between each number's position in start state and its position in goal state.

- Can we make do with inadmissible heuristics?

Yes—example coming up in next section. But try to avoid.

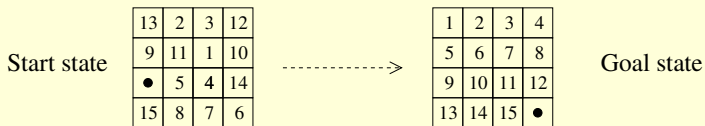
- Is  $A^*$  search used widely in practice?

## Questions

- How to come up with an effective admissible heuristic for a task?

For many tasks people have already done so. A general strategy is to solve the task with relaxed constraints.

- What's a good heuristic for 15-puzzle?



Sum of Manhattan distances between each number's position in start state and its position in goal state.

- Can we make do with inadmissible heuristics?

Yes—example coming up in next section. But try to avoid.

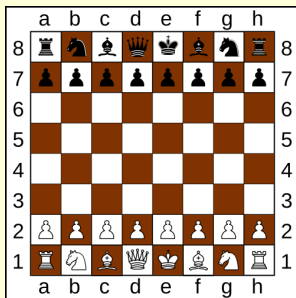
- Is  $A^*$  search used widely in practice?

Yes. Along with variants such as  $IDA^*$ .

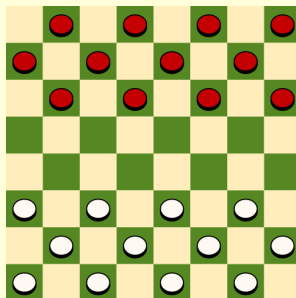
# This Lecture

- Informed search (a.k.a. heuristic search)
- Application in game-playing

# Search and Games



[1]

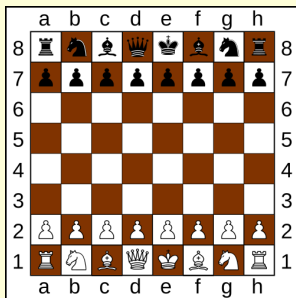


[2]

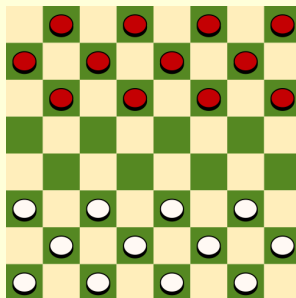
[1] [https://commons.wikimedia.org/wiki/File:AAA\\_SVG\\_Chessboard\\_and\\_chess\\_pieces\\_02.svg](https://commons.wikimedia.org/wiki/File:AAA_SVG_Chessboard_and_chess_pieces_02.svg). CC image courtesy of ILA-boy on Wikimedia Commons licensed under CC-BY-SA-3.0.

[2] <https://commons.wikimedia.org/wiki/File:Draughts.svg>.

# Search and Games



Chess [1]

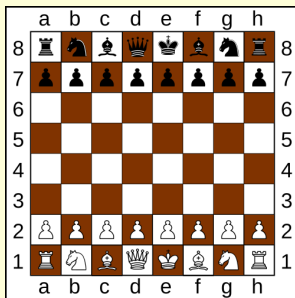


Checkers/Draughts [2]

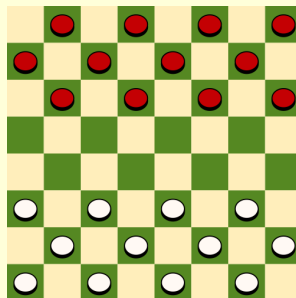
[1] [https://commons.wikimedia.org/wiki/File:AAA\\_SVG\\_Chessboard\\_and\\_chess\\_pieces\\_02.svg](https://commons.wikimedia.org/wiki/File:AAA_SVG_Chessboard_and_chess_pieces_02.svg). CC image courtesy of ILA-boy on Wikimedia Commons licensed under CC-BY-SA-3.0.

[2] <https://commons.wikimedia.org/wiki/File:Draughts.svg>.

# Search and Games



Chess [1]



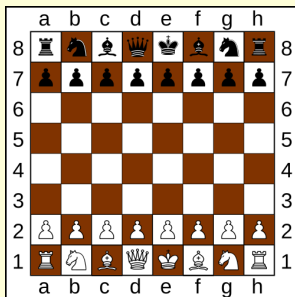
Checkers/Draughts [2]

- Can winning at chess/checkers be posed as a search problem?

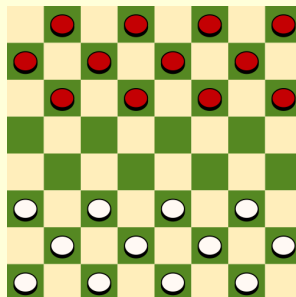
[1] [https://commons.wikimedia.org/wiki/File:AAA\\_SVG\\_Chessboard\\_and\\_chess\\_pieces\\_02.svg](https://commons.wikimedia.org/wiki/File:AAA_SVG_Chessboard_and_chess_pieces_02.svg). CC image courtesy of ILA-boy on Wikimedia Commons licensed under CC-BY-SA-3.0.

[2] <https://commons.wikimedia.org/wiki/File:Draughts.svg>.

# Search and Games



Chess [1]



Checkers/Draughts [2]

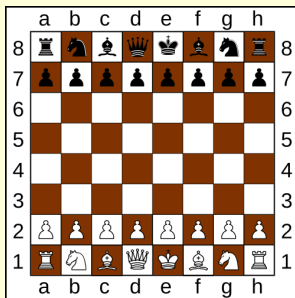
- Can winning at chess/checkers be posed as a search problem?
- What's the main difference from our previous examples?

[1] [https://commons.wikimedia.org/wiki/File:AAA\\_SVG\\_Chessboard\\_and\\_chess\\_pieces\\_02.svg](https://commons.wikimedia.org/wiki/File:AAA_SVG_Chessboard_and_chess_pieces_02.svg). CC image courtesy of ILA-boy on Wikimedia Commons licensed under CC-BY-SA-3.0.

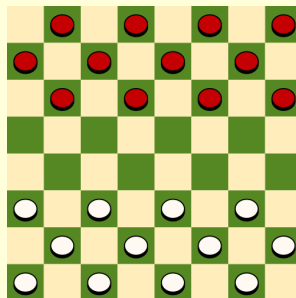
[2] <https://commons.wikimedia.org/wiki/File:Draughts.svg>.



# Search and Games



Chess [1]



Checkers/Draughts [2]

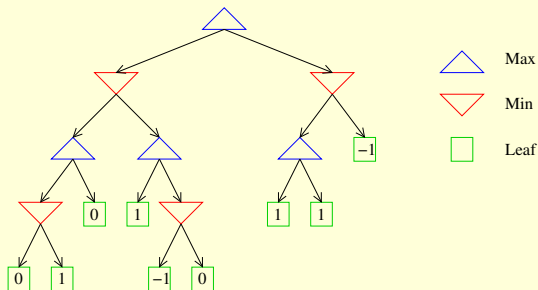
- Can winning at chess/checkers be posed as a search problem?
- What's the main difference from our previous examples? **There's another player!**

[1] [https://commons.wikimedia.org/wiki/File:AAA\\_SVG\\_Chessboard\\_and\\_chess\\_pieces\\_02.svg](https://commons.wikimedia.org/wiki/File:AAA_SVG_Chessboard_and_chess_pieces_02.svg). CC image courtesy of ILA-boy on Wikimedia Commons licensed under CC-BY-SA-3.0.

[2] <https://commons.wikimedia.org/wiki/File:Draughts.svg>.

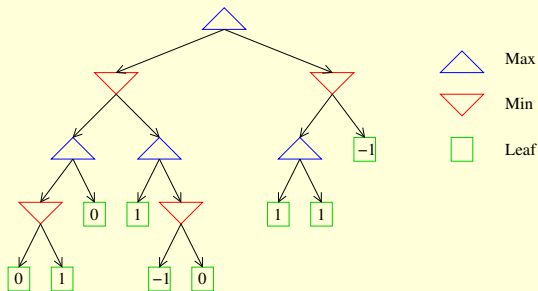
# Game Tree

- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.



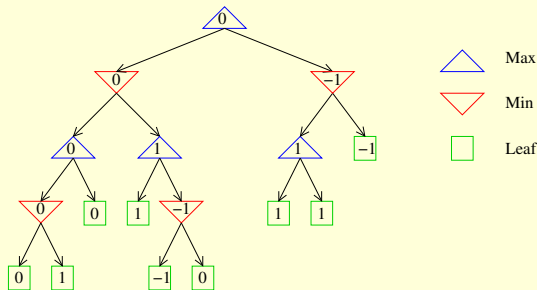
# Game Tree

- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  
 $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.
- What is the value of the root node?**



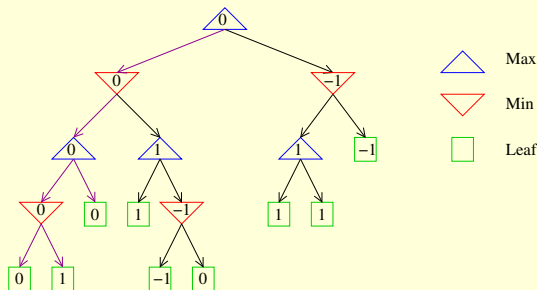
# Game Tree

- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  
 $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.
- What is the value of the root node?**



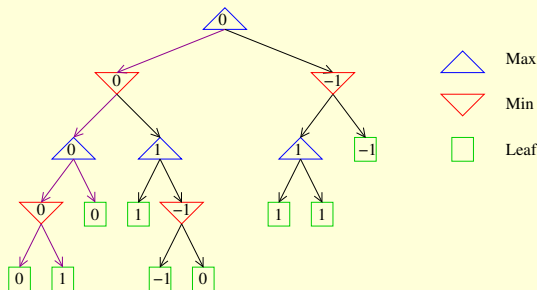
# Game Tree

- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  
 $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.
- What is the value of the root node?**



# Game Tree

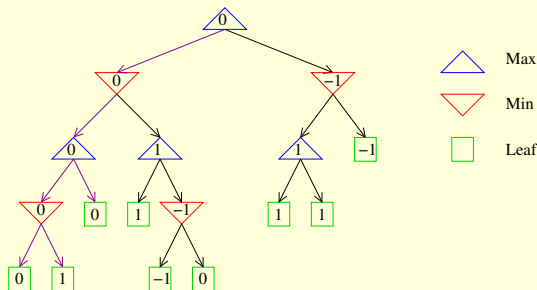
- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  
 $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.
- What is the value of the root node?**



- In 2007, a massive, long-running computation concluded that the value of the root node for Checkers is 0 (draw).

# Game Tree

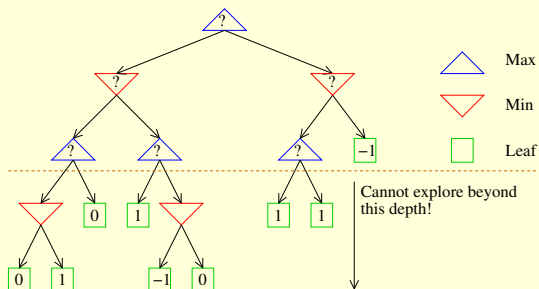
- Assume turn-taking zero sum game with two players, **Max** and **Min**.
- Action costs usually taken as 0, but leaves have value  
 $-1$  (Max loses),  $0$  (draw),  $1$  (Max wins).
- Value of Max node is maximum of values of children.  
Value of Min node is minimum of values of children.
- What is the value of the root node?**



- In 2007, a massive, long-running computation concluded that the value of the root node for Checkers is 0 (draw).
- The Checkers game tree has  $\approx 10^{40}$  nodes; Chess has  $\approx 10^{120}$ .

# Evaluation Function

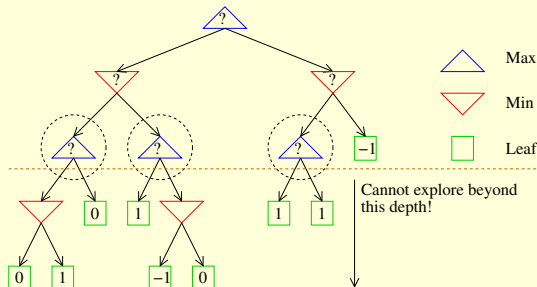
- What if game tree depth/size makes it infeasible to solve?





# Evaluation Function

- What if game tree depth/size makes it infeasible to solve?



- At some depth  $d$  from current node, estimate node value using **features**.
- For example, in Chess, set **evaluation** as

$$w_1 \times \text{Material difference} + w_2 \times \text{King safety} + w_3 \times \text{pawn strength} + \dots$$

- Weights  $w_1, w_2, w_3, \dots$  are tuned or learned from experience.

- Section 3.6, **Artificial Intelligence: Foundations of Computational Agents**, David Poole and Alan Mackworth, Cambridge University Press, 2010. Available on-line at <https://artint.info/html/ArtInt.html>.