

AI - HEURISTIC SEARCH

- ❖ **Heuristic search** refers to a **search** strategy that attempts to optimize a problem by iteratively improving the solution based on a given **heuristic** function or a cost measure.
- ❖ A Heuristic is a technique to solve a problem faster than classic methods, or to find an approximate solution when classic methods cannot.
- ❖ This is a kind of a shortcut as we often trade one of optimality, completeness, accuracy, or precision for speed.

Fig 4.1 First three levels of the tic-tac-toe state space reduced by symmetry

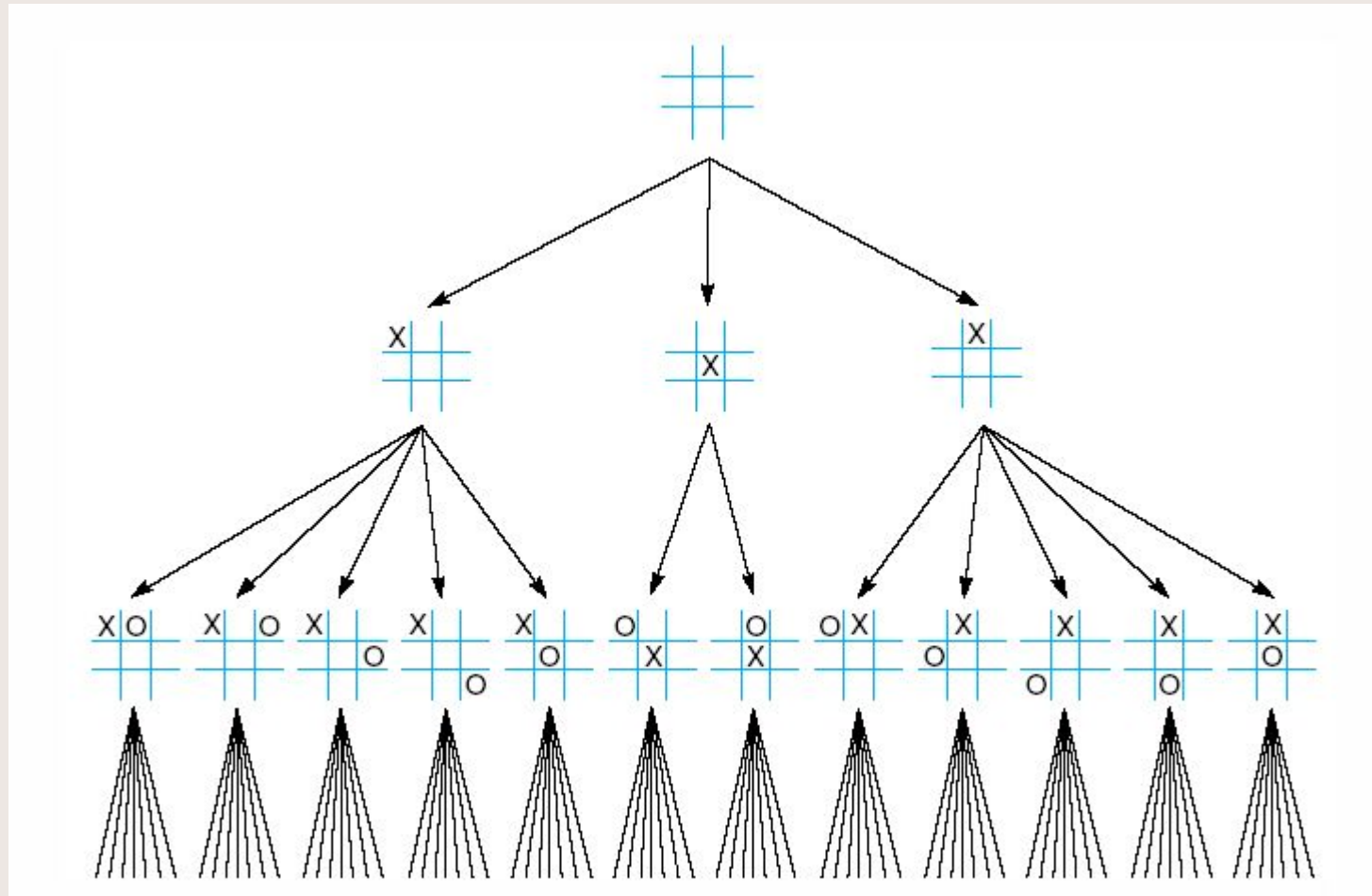
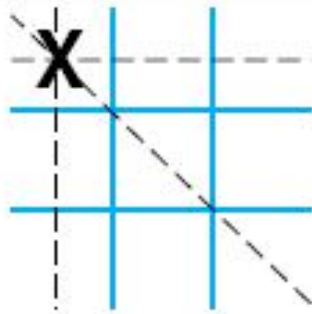
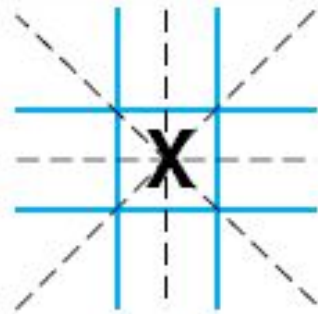


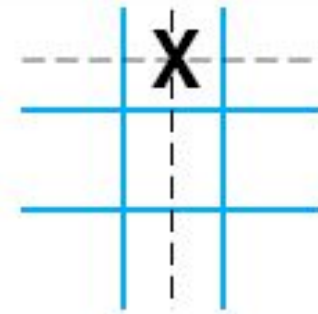
Fig 4.2 The “most wins” heuristic applied to the first children in tic-tac-toe.



Three wins through
a corner square



Four wins through
the center square



Two wins through
a side square

Fig 4.3 Heuristically reduced state space for tic-tac-toe.

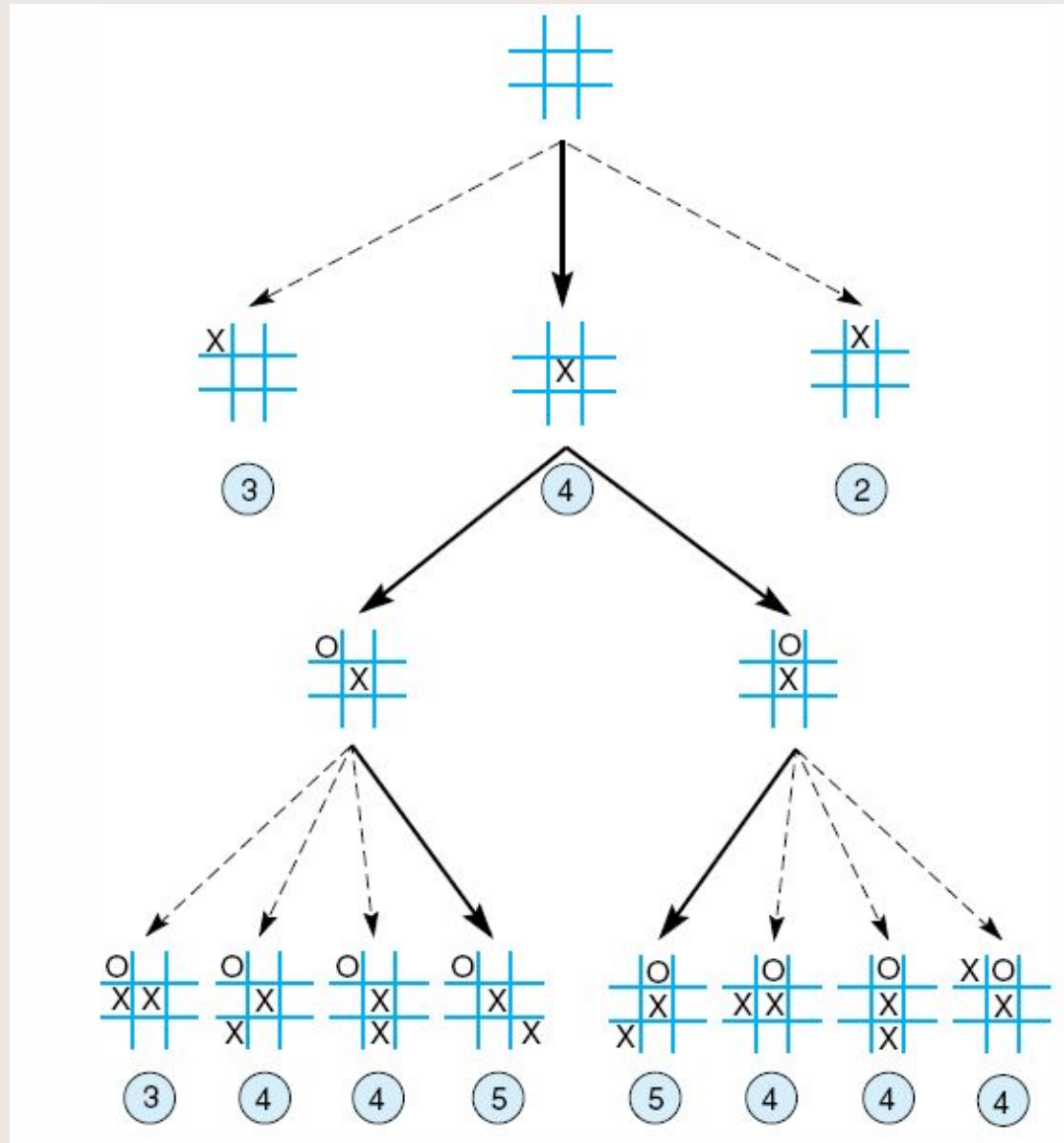


Fig 4.12 The start state, first moves, and goal state for an example-8 puzzle.

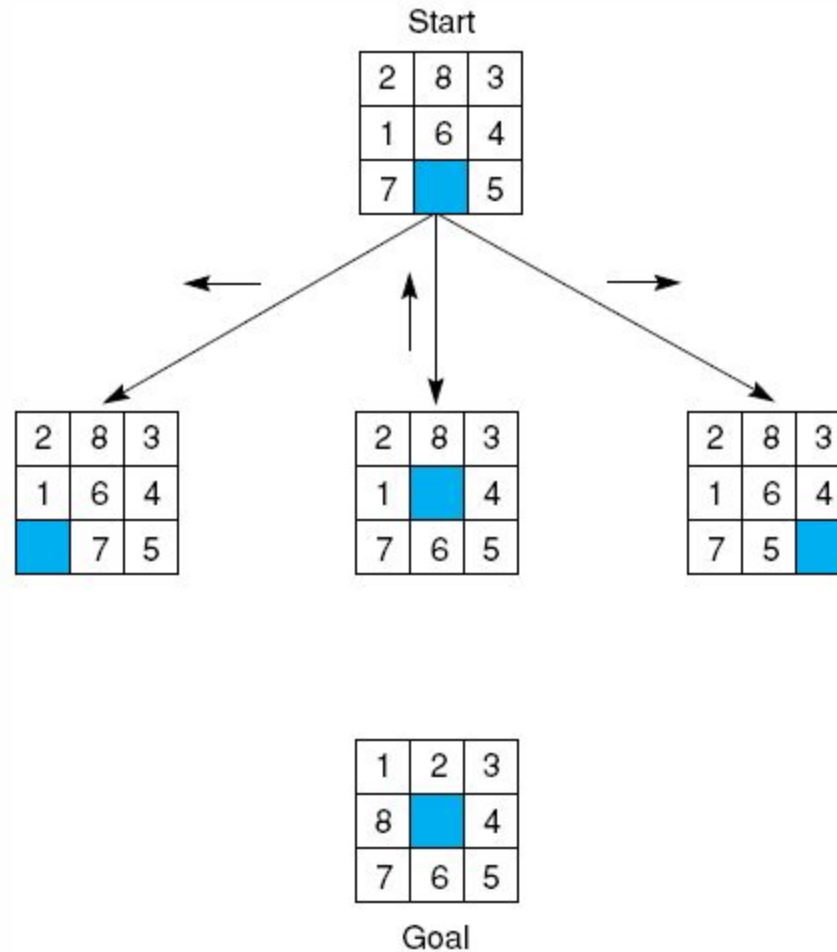


Fig 4.14 Three heuristics applied to states in the 8-puzzle.

<table><tr><td>2</td><td>8</td><td>3</td></tr><tr><td>1</td><td>6</td><td>4</td></tr><tr><td>7</td><td>7</td><td>5</td></tr></table>	2	8	3	1	6	4	7	7	5	5	6	0
2	8	3										
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2	8	3										
1	6	4										
7	5	6										
	Tiles out of place	Sum of distances out of place	2 x the number of direct tile reversals									

1	2	3
8	6	4
7	6	5

Goal

Fig 4.15 The heuristic f applied to states in the 8-puzzle.

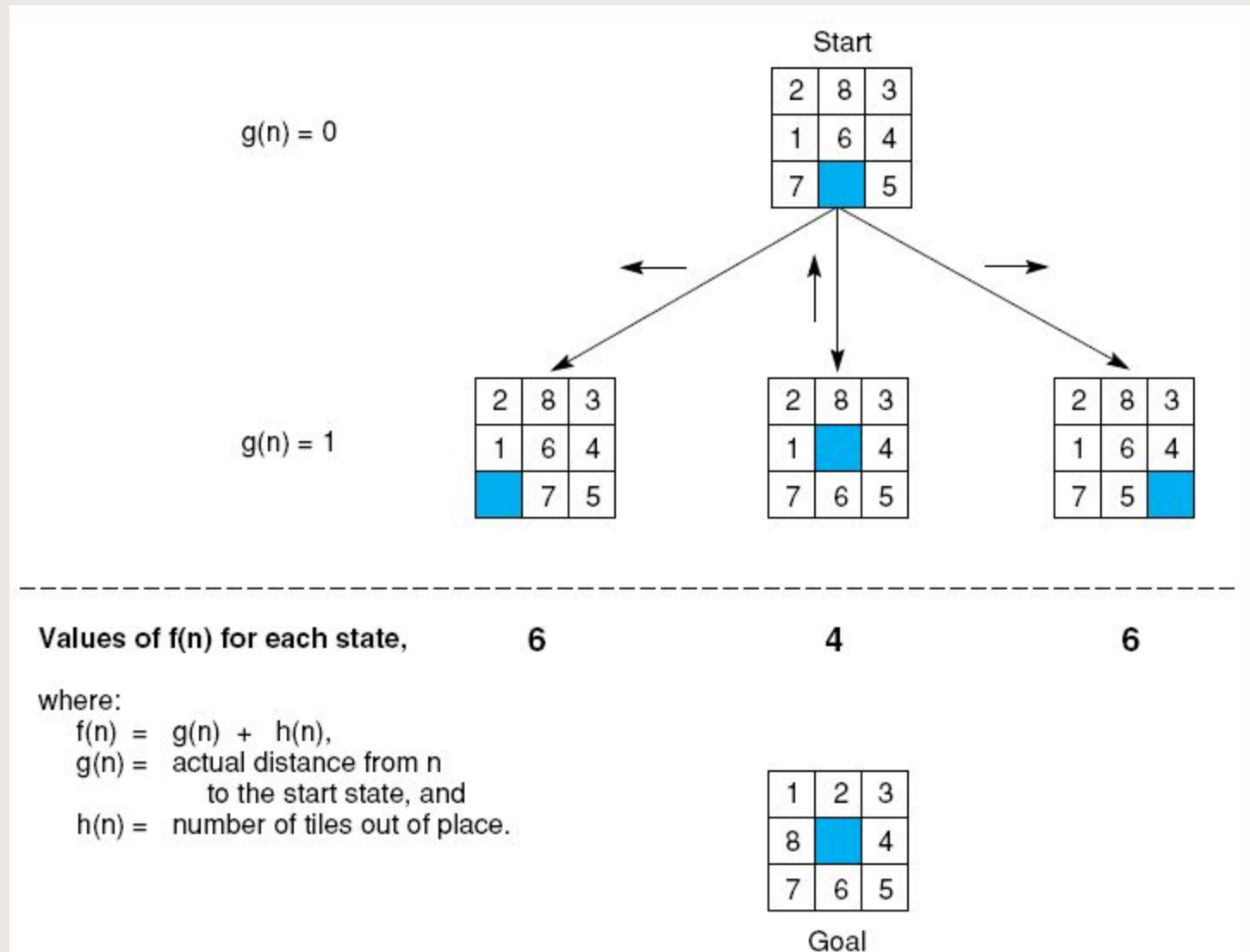


Fig 4.16 State space generated in heuristic search of the 8-puzzle graph.

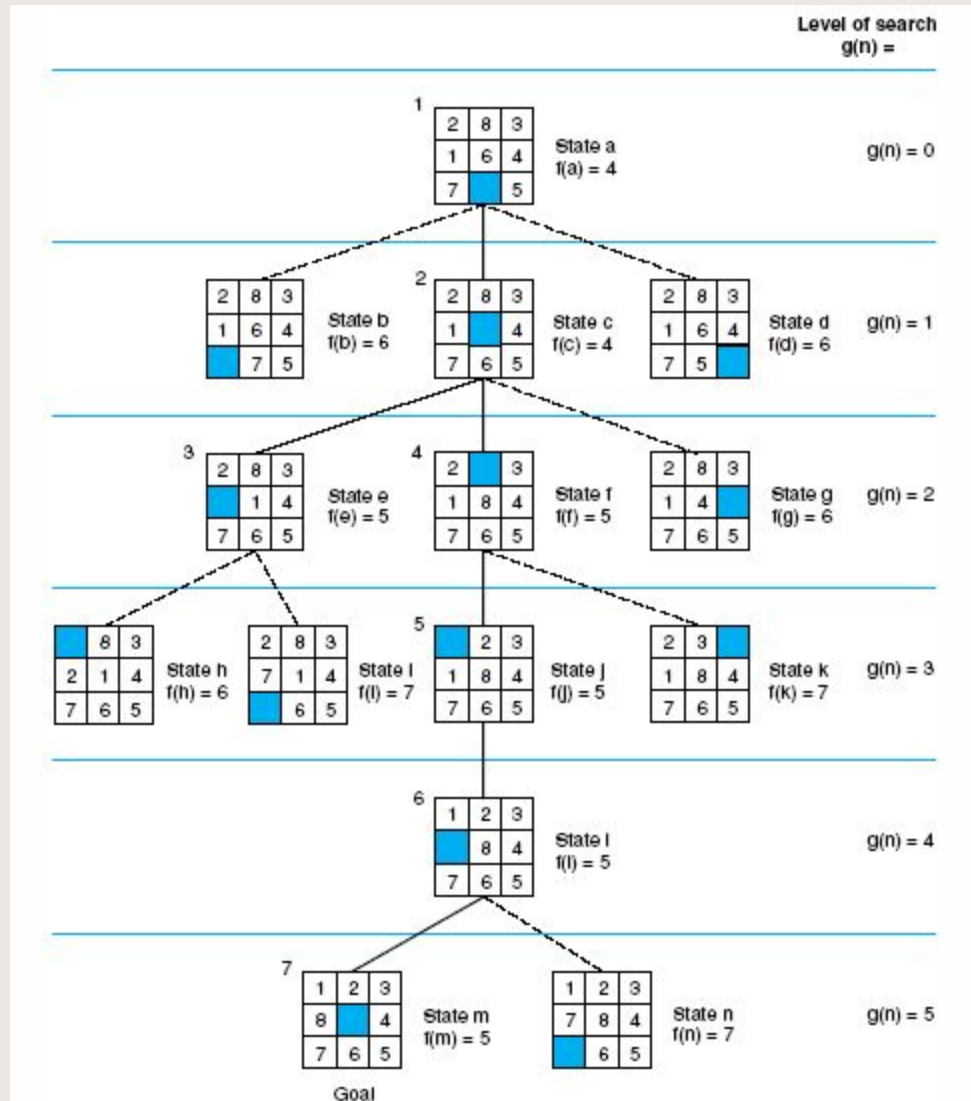
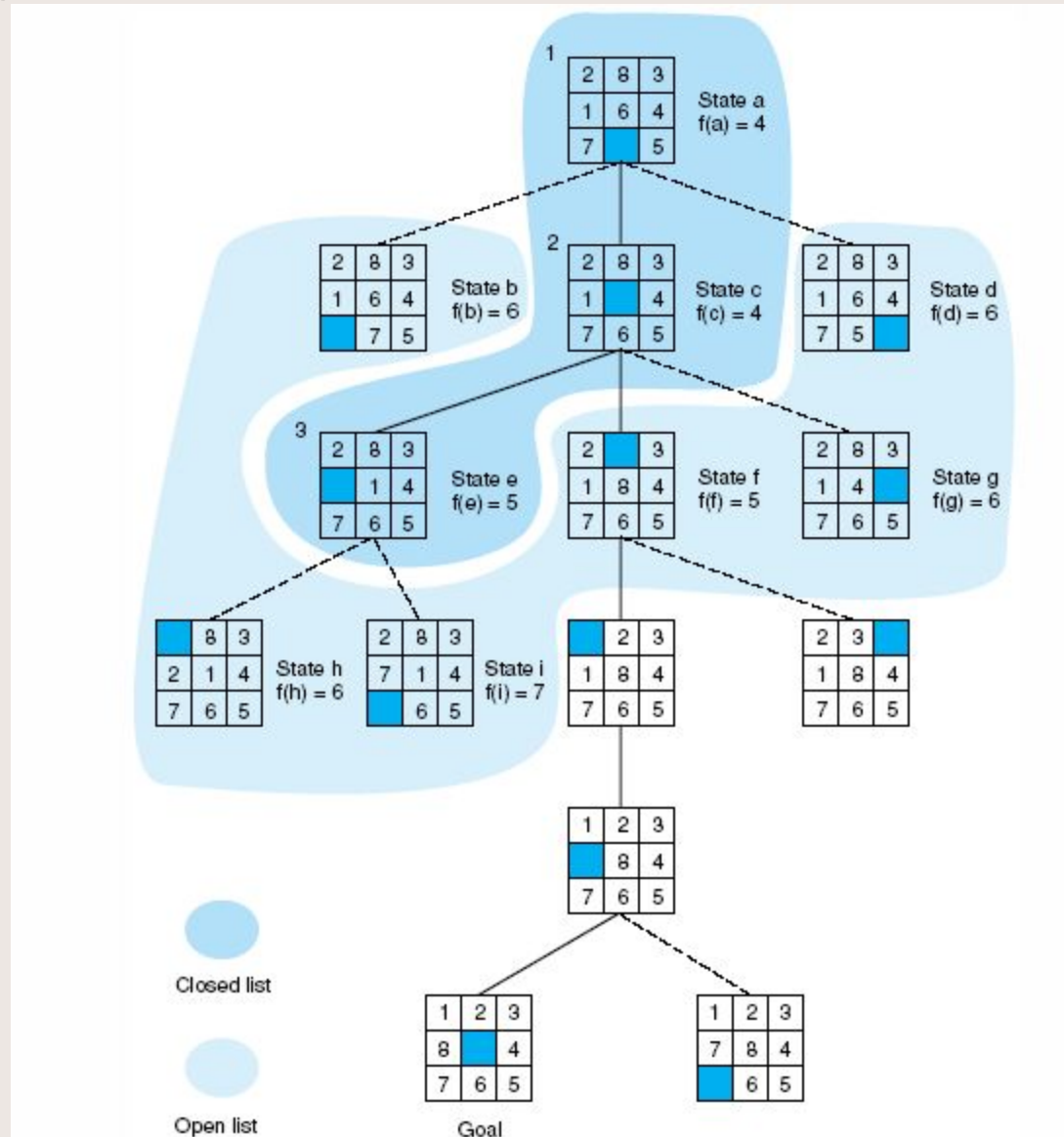


Fig 4.17 Open and closed as they appear after the 3rd iteration of heuristic search



DEFINITION

ALGORITHM A, ADMISSIBILITY, ALGORITHM A*

Consider the evaluation function $f(n) = g(n) + h(n)$, where

n is any state encountered in the search.

$g(n)$ is the cost of n from the start state.

$h(n)$ is the heuristic estimate of the cost of going from n to a goal.

If this evaluation function is used with the **best_first_search** algorithm of Section 4.1, the result is called *algorithm A*.

A search algorithm is *admissible* if, for any graph, it always terminates in the optimal solution path whenever a path from the start to a goal state exists.

If algorithm A is used with an evaluation function in which $h(n)$ is less than or equal to the cost of the minimal path from n to the goal, the resulting search algorithm is called *algorithm A** (pronounced “A STAR”).

It is now possible to state a property of **A*** algorithms:

All **A*** algorithms are admissible.

DEFINITION

MONOTONICITY

A heuristic function h is monotone if

1. For all states n_i and n_j , where n_j is a descendant of n_i ,

$$h(n_i) - h(n_j) \leq \text{cost}(n_i, n_j),$$

where $\text{cost}(n_i, n_j)$ is the actual cost (in number of moves) of going from state n_i to n_j .

2. The heuristic evaluation of the goal state is zero, or $h(\text{Goal}) = 0$.

DEFINITION

INFORMEDNESS

For two A* heuristics h_1 and h_2 , if $h_1(n) \leq h_2(n)$, for all states n in the search space, heuristic h_2 is said to be *more informed* than h_1 .

Fig 4.18 Comparison of state space searched using heuristic search with space searched by breadth-first search. The proportion of the graph searched heuristically is shaded. The optimal search selection is in bold. Heuristic used is $f(n) = g(n) + h(n)$ where $h(n)$ is tiles out of place.

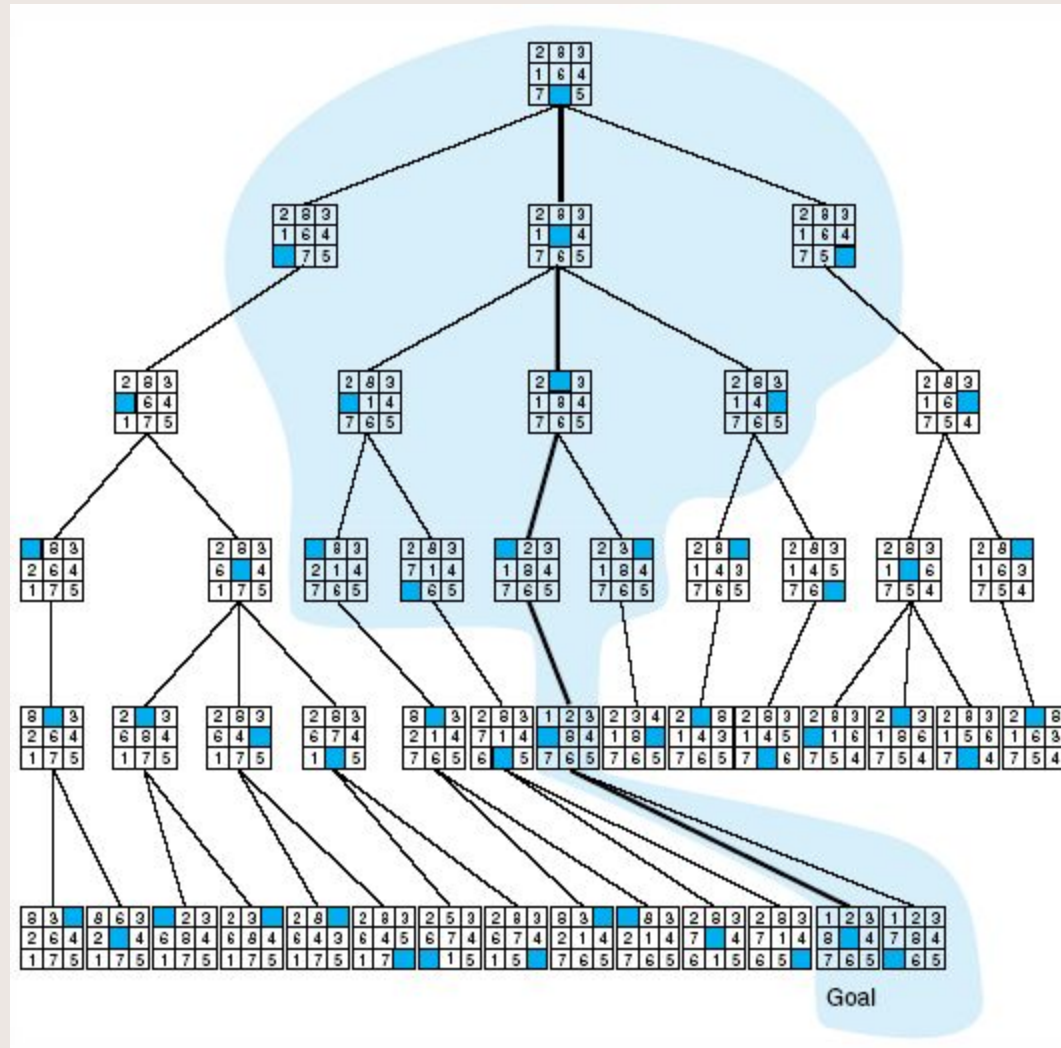


Fig 4.19 State space for a variant of nim. Each state partitions the seven matches into one or more piles.

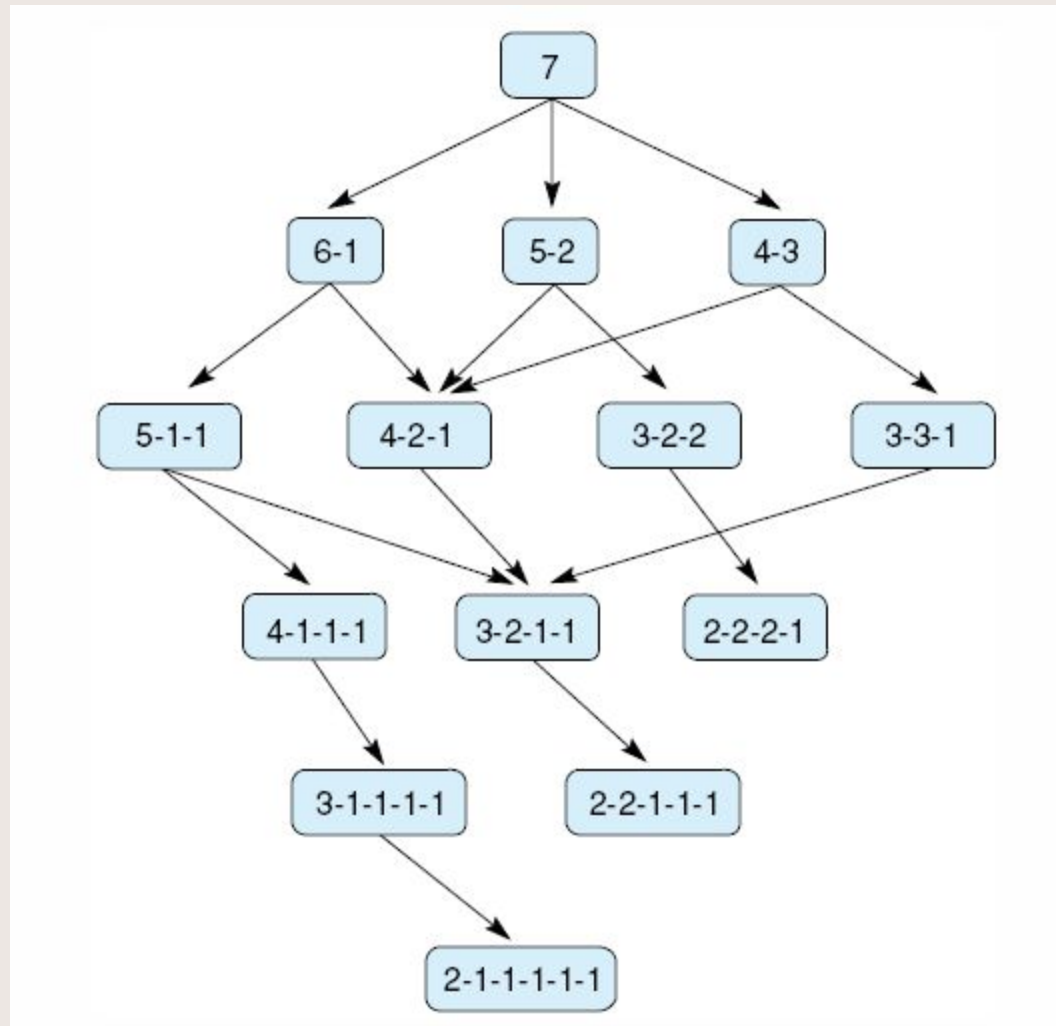
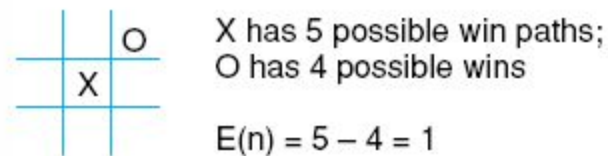
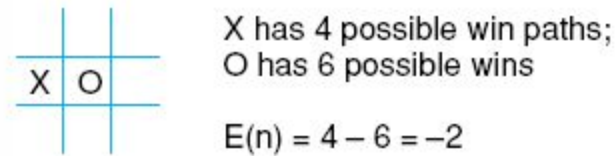
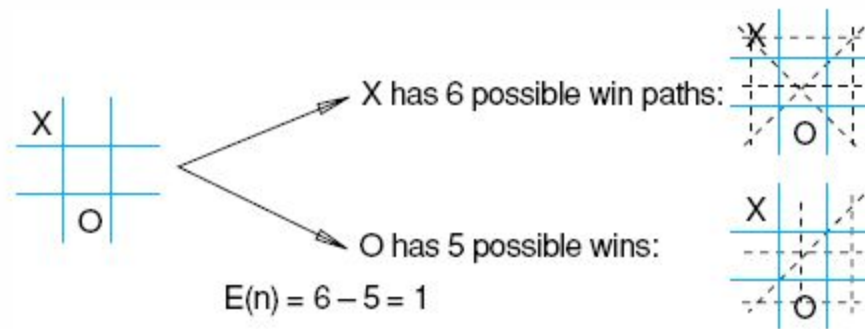


Fig 4.22 Heuristic measuring conflict applied to states of tic-tac-toe.



Heuristic is $E(n) = M(n) - O(n)$

where $M(n)$ is the total of My possible winning lines

$O(n)$ is total of Opponent's possible winning lines

$E(n)$ is the total Evaluation for state n

Fig 4.23 Two-ply minimax applied to the opening move of tic-tac-toe, from Nilsson (1971).

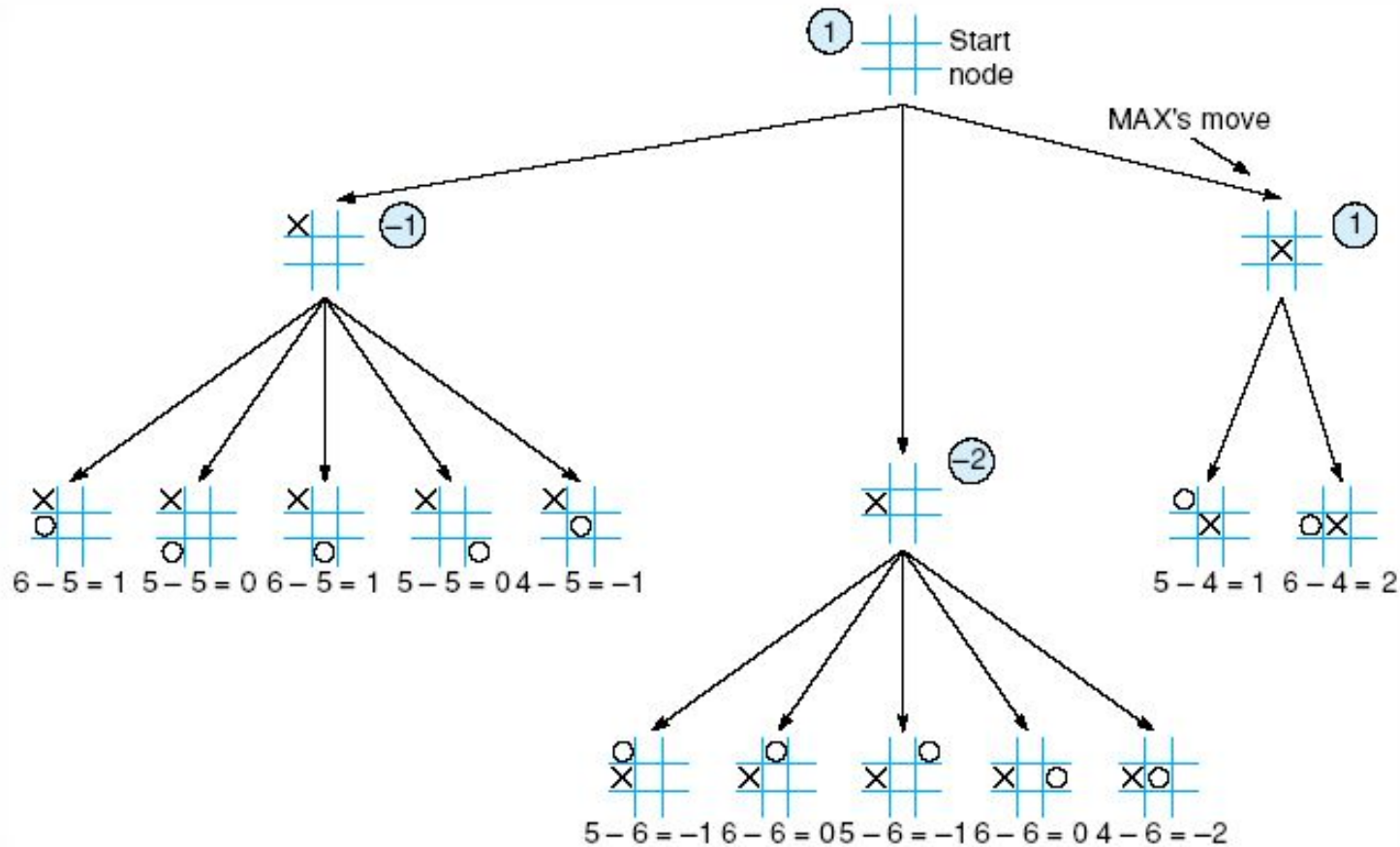


Fig 4.24 Two ply minimax, and one of two possible MAX second moves, from Nilsson (1971).

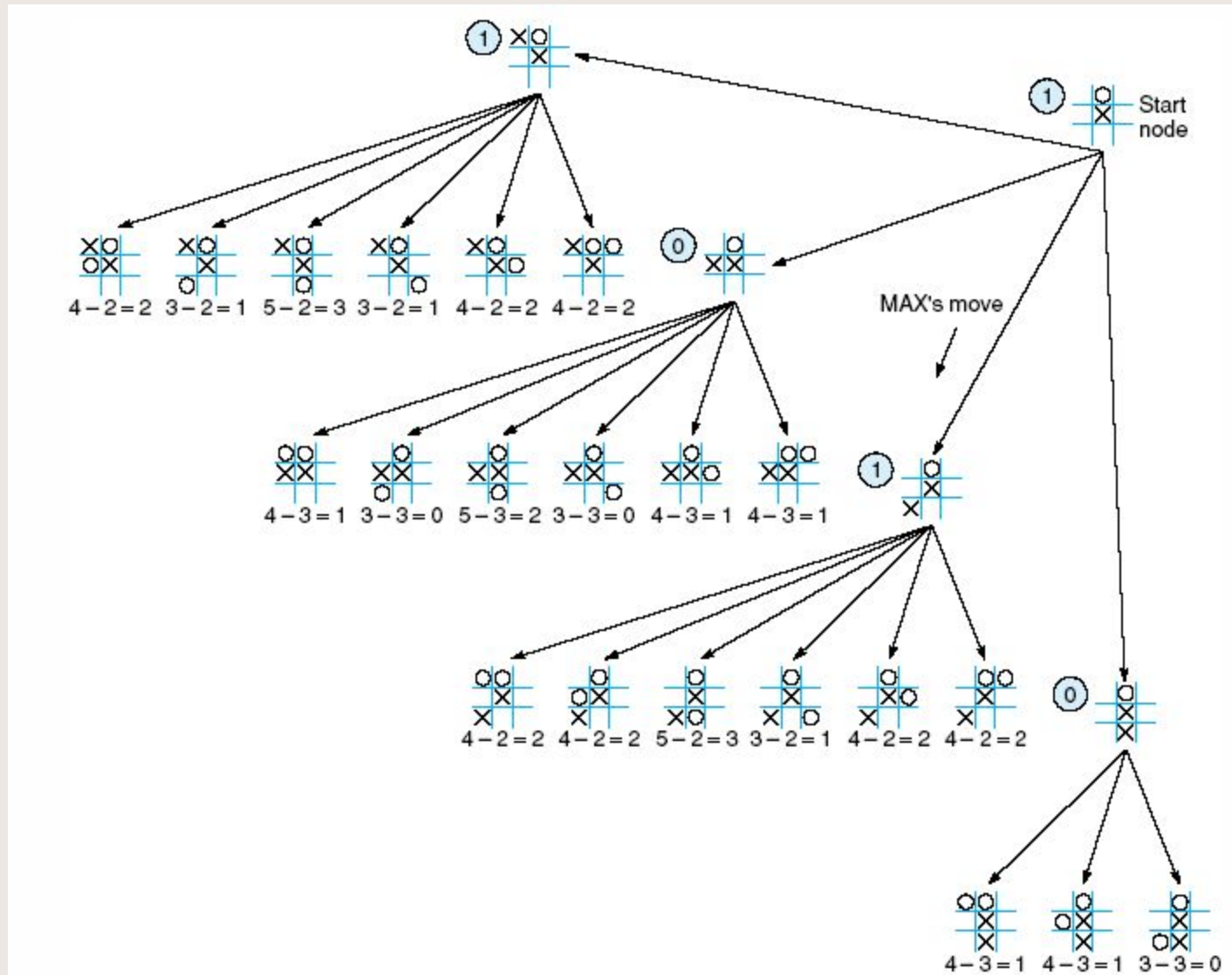


Fig 4.25 Two-ply minimax applied to X's move near the end of the game, from Nilsson (1971).

