



# COMP 3200

## Artificial Intelligence

### **Lecture 5**

#### Avoiding Repeated States

#### Heuristic Search

# Note: Terminology

- Algorithm terminology / name from
- Artificial Intelligence: A Modern Approach
  - Stuart Russel and Peter Norvig
  - Not required for course, but amazing book
- Names not important, concepts are

# Avoiding Repeated States

- One of the **most important** search ideas
- Especially important with reversible actions
- Most infinite loops / wasted time can be avoided by not returning to identical states
- We can remember nodes expanded
  - Don't re-expand them
- Can lead to **exponential savings** in the number of nodes generated

# 'Closed List'

- Store states we have already expanded
- **Closed List** stores expanded states
  - Often implemented as a hash table / dictionary for efficient lookup
- Check if a node is closed before expanding
- **Open List** = fringe of unexpanded nodes
  - Often implemented as a priority queue

# General Tree-Search (recall)

1. Function **Graph-Search**(problem, strategy)
2.     open = {Node(problem.initial\_state)}
3.     **while** (true)
4.         **if** (open.empty) **return** fail
5.         node = strategy.select\_node(open)
6.         **if** (node.state is goal) **return** solution
7.         open.add(Expand(node, problem))

# General Graph-Search

1. Function **Graph-Search**(problem, strategy)
2.     `closed = {} // add closed list`
3.     `open = {Node(problem.initial_state)}`
4.     **while** (true)
5.         **if** (open.empty) **return** fail
6.         `node = strategy.select_node(open)`
7.         **if** (node.state is goal) **return** solution
8.         `if (node.state in closed) continue // check if closed`
9.         `closed.add(node.state) // mark node closed`
10.        `open.add(Expand(node, problem))`

# General Graph-Search

1. Function **Graph-Search**(problem, strategy)
2.     closed = { }
3.     open = { Node(problem.initial\_state) }
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# Assignment 1: BFS

1. Function **Graph-Search**(problem, BFS)
2.     closed = { }
3.     open = Queue{ Node(problem.initial\_state) }
4.     **while** (true)
5.         **if** (open.empty) **return** fail
6.         node = open.pop() // get from front of queue
7.         **if** (node.state is goal) **return** solution
8.         **if** (node.state in closed) **continue**
9.         closed.add(node.state)
10.        open.add(Expand(node, problem))



# Assignment 1: DFS

1. Function **Graph-Search**(problem, DFS)
2.     closed = { }
3.     open = Stack{Node(problem.initial\_state)}
4.     **while** (true)
5.         **if** (open.empty) **return** fail
6.         node = open.pop() // get from top of stack
7.         **if** (node.state is goal) **return** solution
8.         **if** (node.state in closed) **continue**
9.         closed.add(node.state)
10.        open.add(Expand(node, problem))

# Tree-Search vs Graph-Search

- Tree Search remembers nothing
  - Will re-expand states multiple times
- Graph-Search remembers states visited
  - Will never re-expand state a second time
- Tree/Graph Search = Just a name

# Graph-Search Complexity

- **Time Complexity**  $O(\text{states})$ 
  - Each state expanded at most once
  - Worst case, expand every state in environment
  - In most cases, far lower than  $O(b^d)$
  - Same upper bound no matter which search strategy
- **Space Complexity**  $O(\text{states})$ 
  - Must store entire search space in memory
  - Space = nodes generated = # states
  - Better, but often times still too much
  - Previously linear space methods may now use more memory

# Graph-Search Optimality

- **Graph-Search may no longer be Optimal**
  - Tricky issue, since it depends on underlying strategy
  - When a repeated state is detected, the algorithm has found another path to the same state
  - Graph-Search (as shown) keeps the original path, and discards new paths found, which may be optimal
  - Only optimal if we can guarantee that the first solution found is the optimal one (BFS, UCS, with same costs)

# Bonus Enhancement

- Open list nodes contain their g values
- When generating a node, check its g value
- If a node already exists on the open list with a  $\leq$  g value, it has a shorter path
- The path we generated has  $> g$ , so is worse
- Do not add the generated node to open list
- Note: Only for action costs  $> 0$

# Enhanced Graph-Search

```
1. Function Graph-Search(problem, strategy)
2.     closed = {}
3.     open = {Node(problem.initial_state)}
4.     while (true)
5.         if (open.empty) return fail
6.         node = strategy.select_node(open)
7.         if (node.state is goal) return solution
8.         if (node.state in closed) continue
9.         closed.add(node.state)
10.        for c in Expand(node, problem)
11.            if (node n in open with c.state and  $n.g \leq c.g$ ) continue
12.            open.add(c)
```

# Enhanced Graph-Search

```
1. Function Graph-Search(problem, strategy)
2.     closed = {}
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10.        for c in Expand(node, problem)
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12.            open.add(c)
```

# Heuristic Search



# Informed (Heuristic) Search

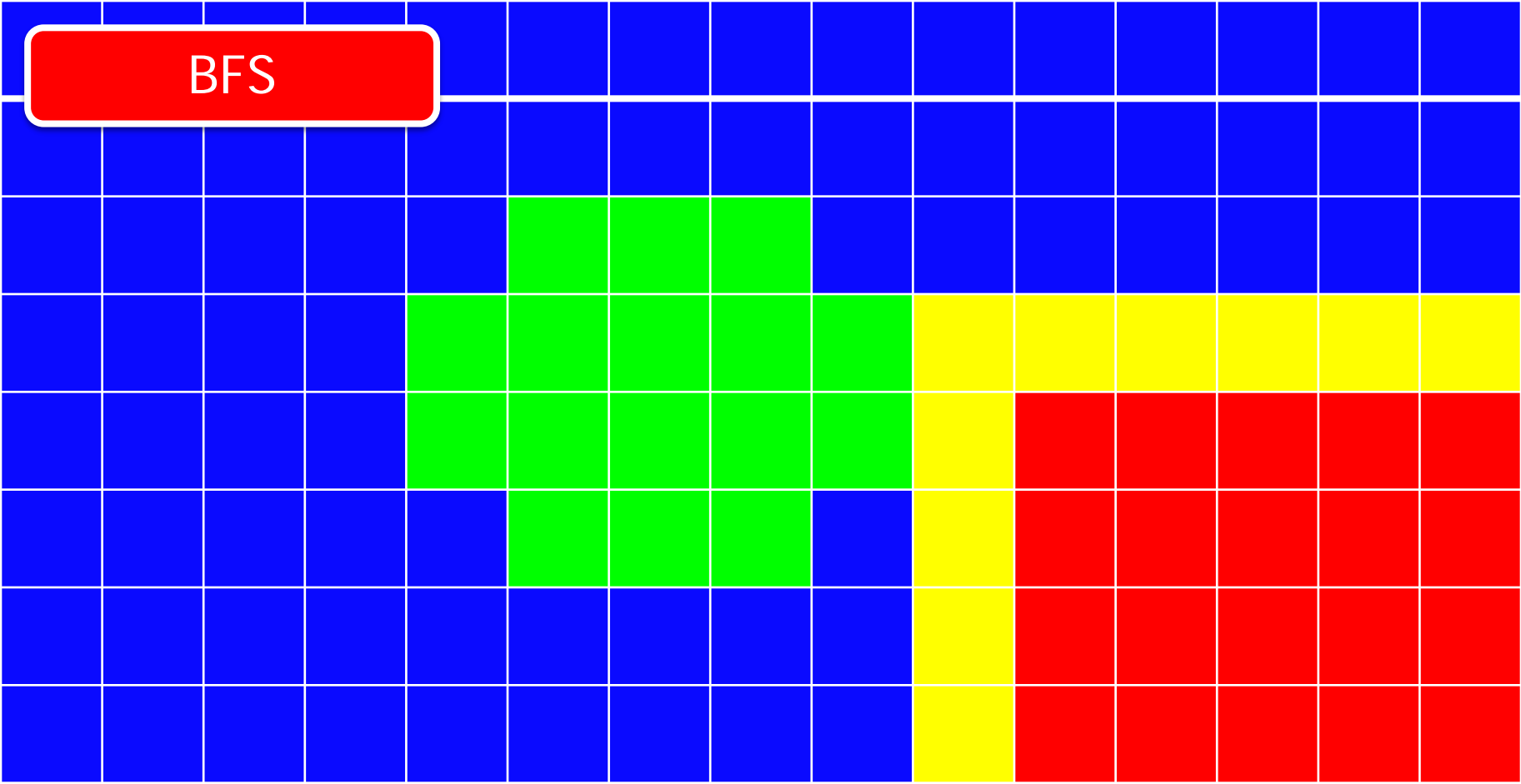
- An informed search strategy uses **problem-specific knowledge** beyond just the problem description itself
- Uses guesses (**heuristics**) to guide search toward the direction of the goal
- Reduce total nodes searched
- Speeds up search times to goal

# BFS

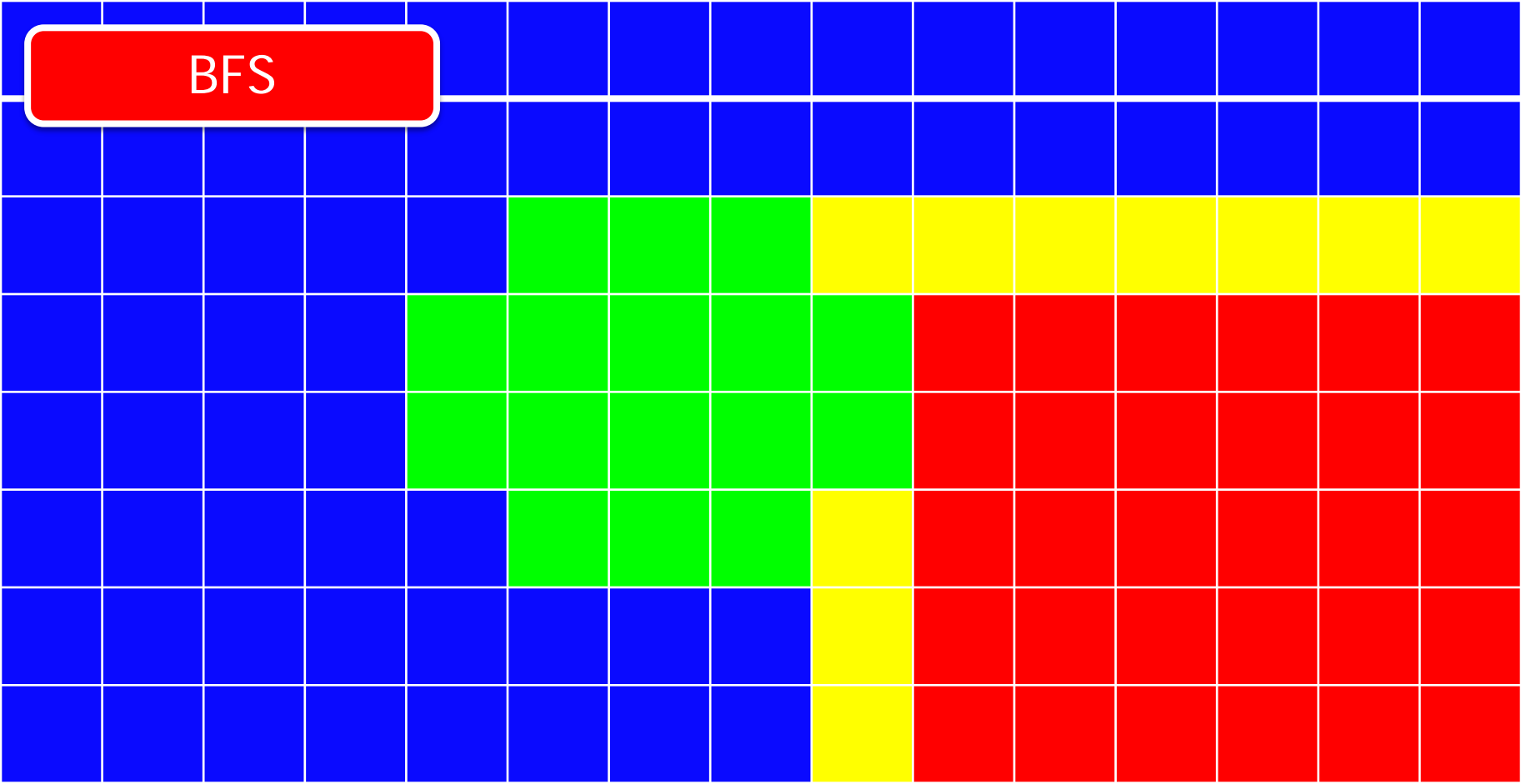
BFS

# BFS

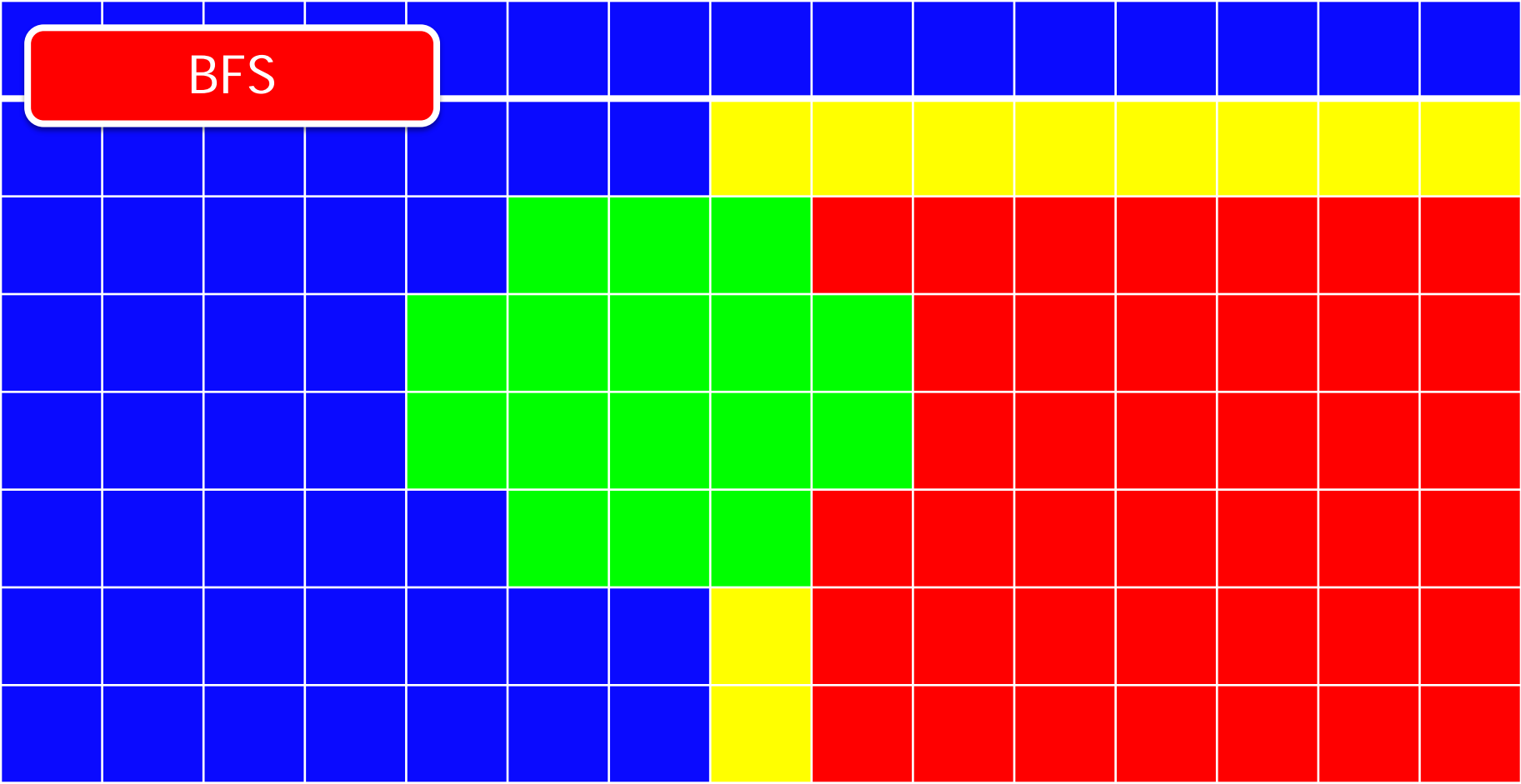
BFS



BFS



BFS



# Best-First Search (BeFS)

- Instance of general tree search
- Select a node for expansion based on an **evaluation function** =  $f(n)$
- Select the node with **minimum value** of  $f(n)$ , measures distance to goal
- “Best First” is slightly misleading
  - Knowing the true best node = go straight to goal
  - More like “Best Guess First”



# Best-First Search (BeFS)

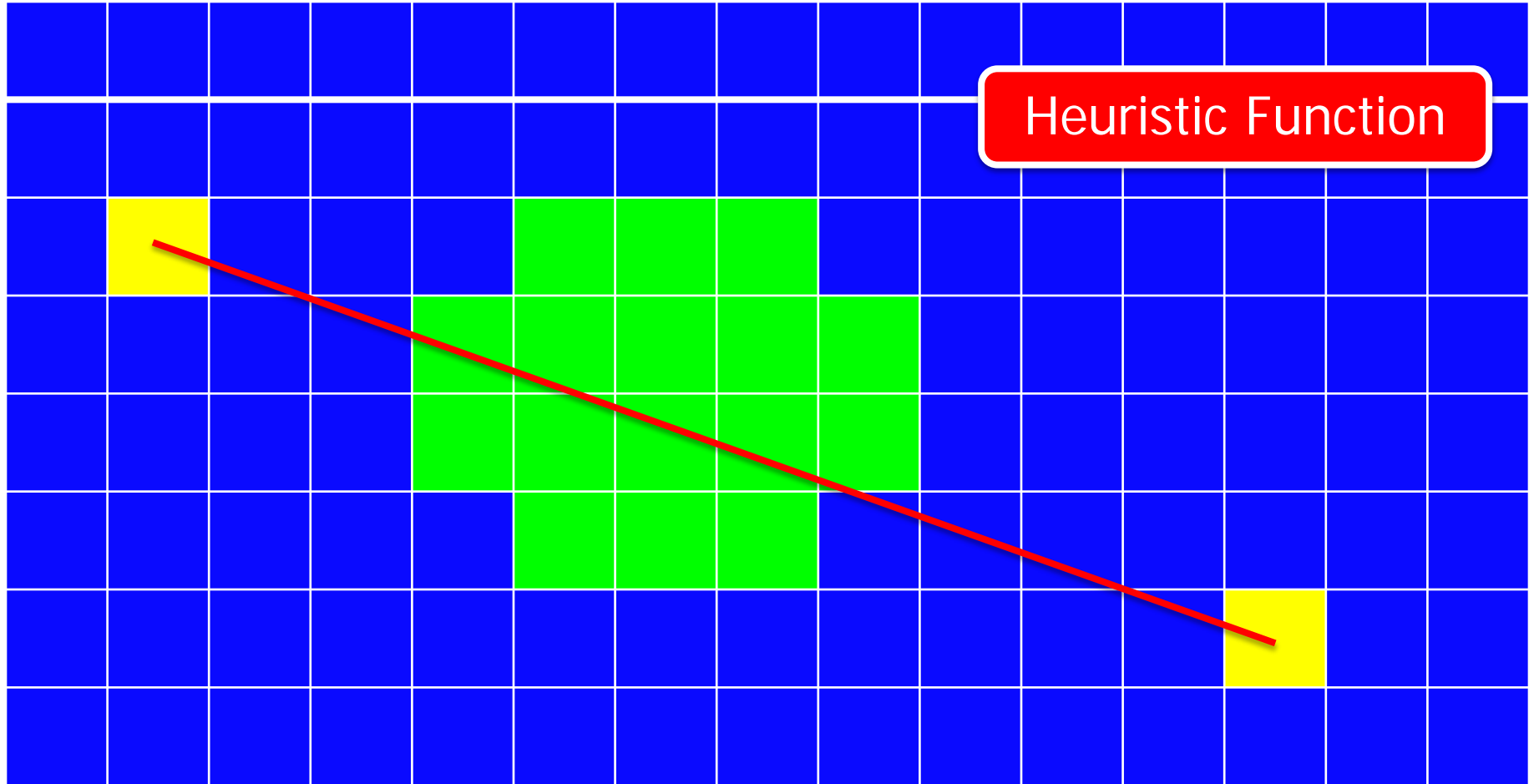
- BeFS can be implemented with general Tree-Search by using a **priority queue** for the open list, sorted on  $f(n)$
- Whole family of BeFS algorithms
  - Have different evaluation functions
  - Most have a heuristic function  **$h(n)$**

# Heuristic Function $h(n)$

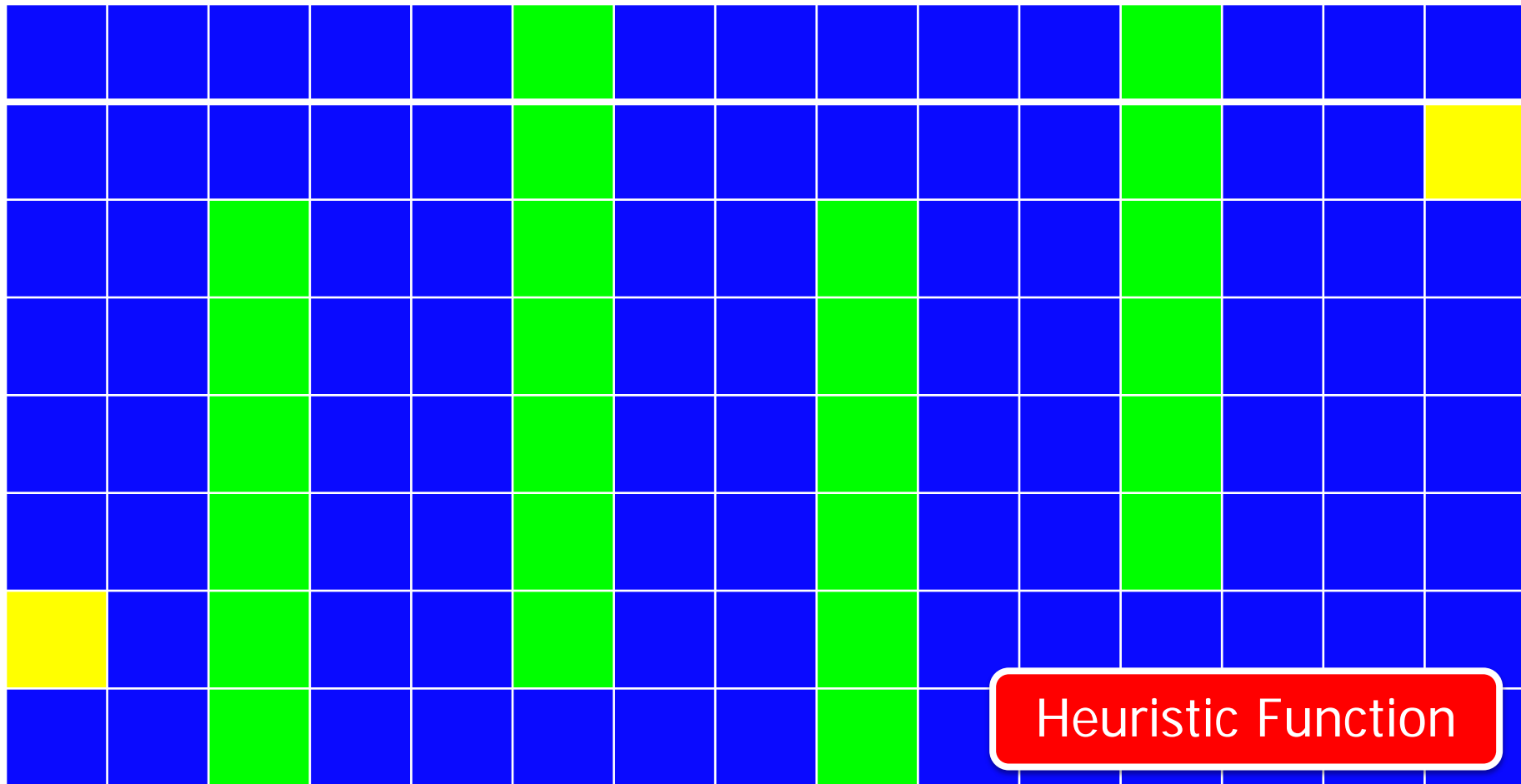
- **Estimated** cost of the **optimal path** from node  $n$  to a goal node
- Heuristic functions are the most common way that additional knowledge of a problem is given to the search algorithm
- If  $n$  is a goal node,  $h(n) = 0$
- Perfect heuristic =  $h^*(n)$

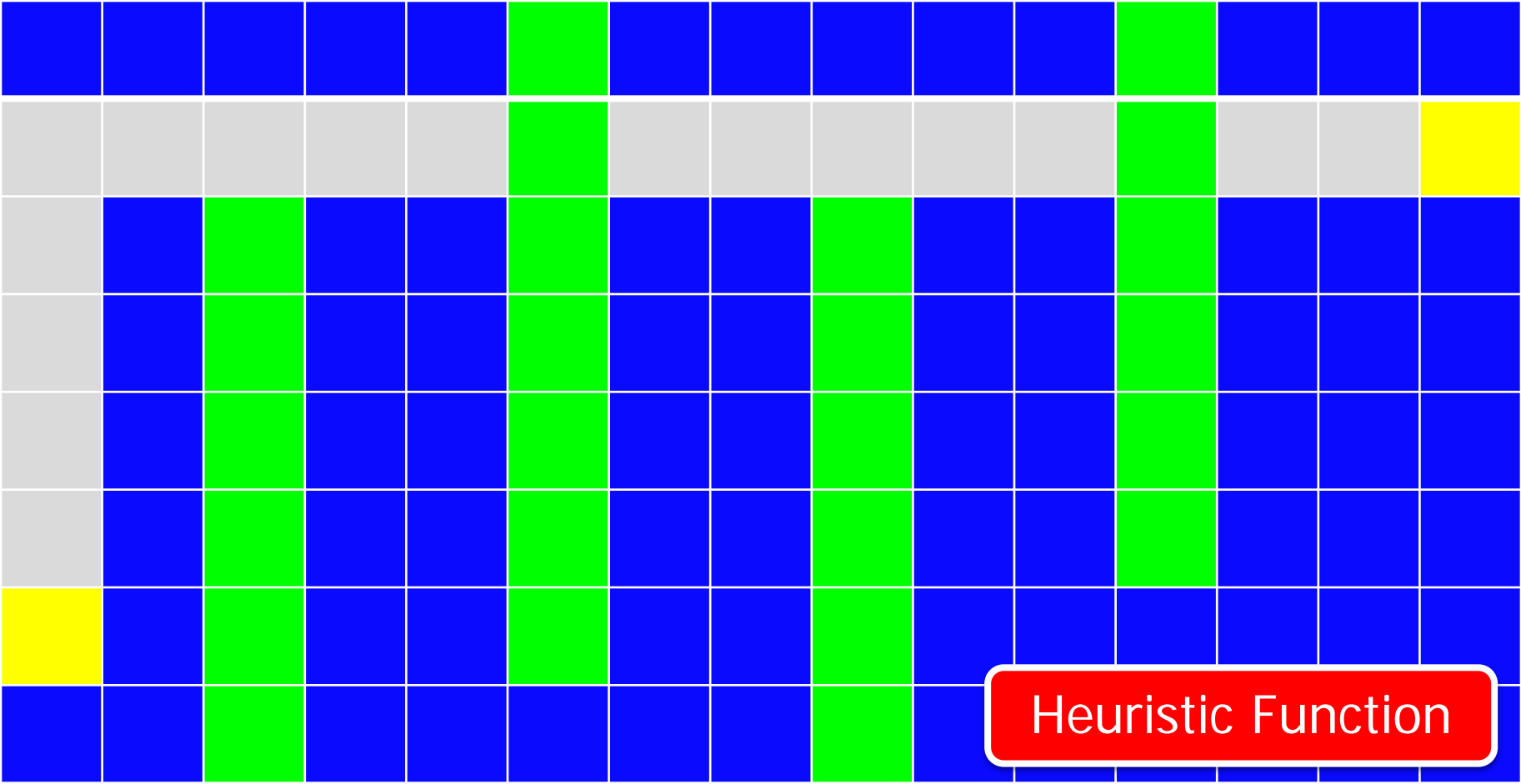
# Heuristic Function

## Heuristic Function



# Heuristic Function





# Greedy Best-First Search

- Expands the node on the open list that it thinks is closest to the goal node
- This will hopefully lead us to the goal
- Thus, for **GBeFS**:  $f(n) = h(n)$
- Resembles DFS
  - Tries a single path all the way to a goal
  - Backs up when it hits a 'dead-end'

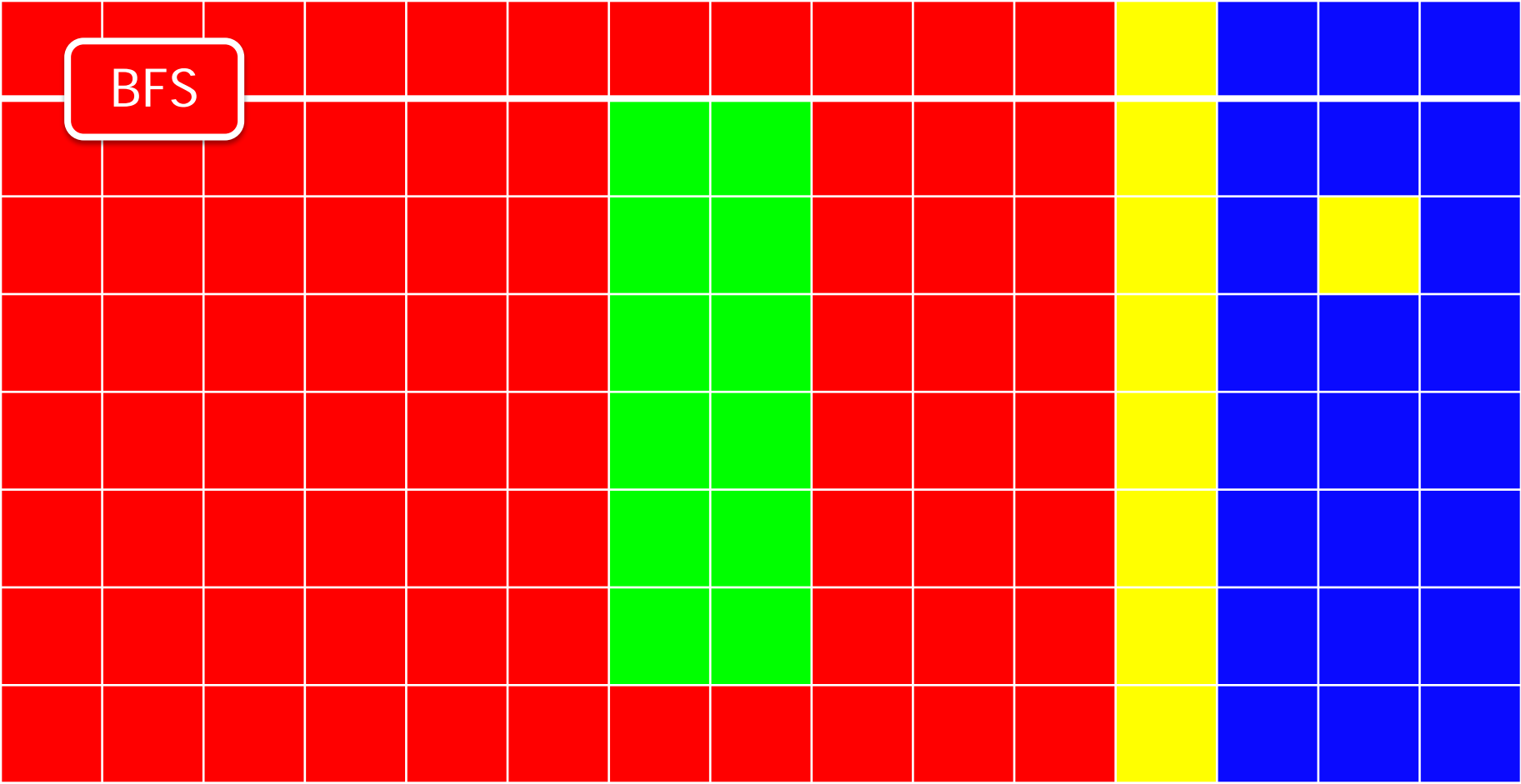


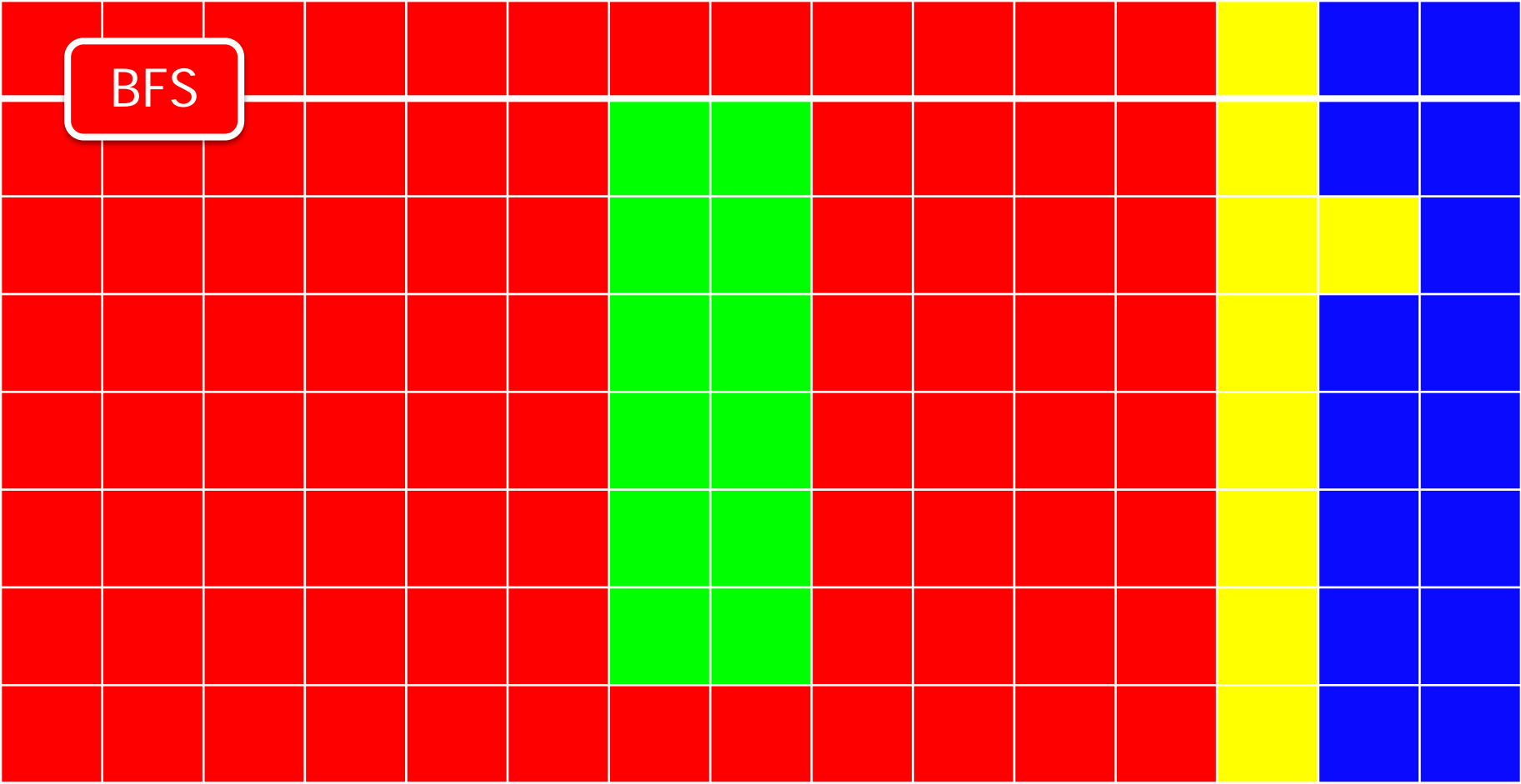






















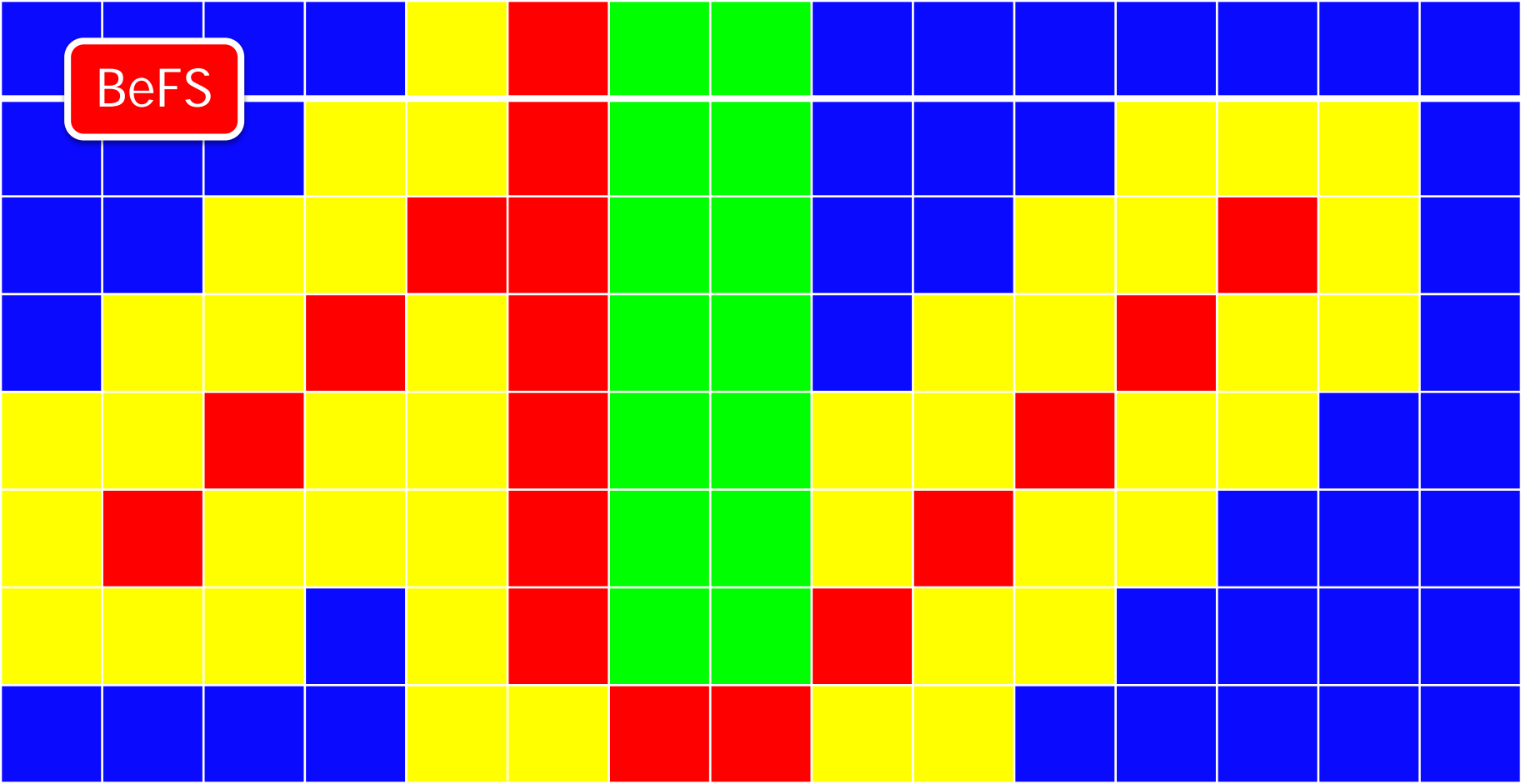








BeFS



# GBeFS Performance

- Suffers similarly to DFS
- Incomplete in General
  - May not find a goal
  - May get lost in paths if heuristic is bad
- Not Optimal
  - May find a higher cost path than optimal
- Time complexity  $O(b^m)$ ,  $m = \text{max depth}$



# Admissible Heuristic

- **Never overestimates** distance to goal
- Can be seen as 'optimistic' guesses
- $h(n) \leq h^*(n)$
- $f(n) = g(n) + h(n)$  never overestimates true cost of a path through  $n$  when  $h(n)$  is admissible

# Heuristic Function

Admissible?

# Admissible?



# Consistent Heuristic

- Also called monotone heuristic
- Consistent if  $h(n) \leq c(n,a,n') + h(n')$ 
  - $h(n)$  = estimate path cost from  $n$  to goal
  - $c(n,a,n')$  = cost of action  $a$  (transitions node  $n$  to  $n'$ )
  - $h(n')$  = estimate path cost from  $n'$  to goal
- Estimate of reaching goal from  $n$  is always less than the estimate of reaching the goal from  $n'$  plus the cost of getting to  $n'$  from  $n$

# Heuristic Function

# Consistent?

# Consistent Heuristic

- Every consistent heuristic is also admissible
- In practice, it is hard to construct an admissible heuristic that isn't consistent
- If  $h(n)$  is consistent, the values of  $f(n)$  along any path are non-decreasing
- The sequence of nodes expanded by  $A^*$  using Graph-Search is in non-decreasing order of  $f(n)$
- Therefore, first goal node selected for expansion is optimal, since all later nodes are at least as expensive
- $A^*$  using Graph-Search is optimal if  $h(n)$  is consistent