



COMP 4752

Computational Intelligence

Lecture 9

Heuristic Search Improvements
Depth-First Branch and Bound
Starcraft AI

A* Search – Memory

- A* search stores nodes in the search tree, so the memory required may far exceed the number of states in the problem
- Worst-Case Memory Required: $O(b^m)$
 - Heuristic may guide us to bad parts of tree
- A* not used in practice for large problems
- How can we reduce memory usage?

Iterative Deepening A^* (IDA^*)

- Iterative deepening used in DFS to ensure the search was complete
- Instead of using a depth limit like DFS, IDA^* uses an increasing f-value cutoff
- But memory / CPU usage still high due to storing and sorting the entire open list

Recursive Best-First Search (RBFS)

- Recall DFS can be implemented using the Tree-Search algorithm with a fringe stack
- However, a better implementation of DFS was using a recursive function
- Similarly, we can implement Best-First Search algorithms as recursive functions
- Mimics the operations of standard BeFS, but only uses linear space

Recursive Best-First Search (RBFS)

- Keeps track of the f -value of the best alternative path from any ancestor of the current node
- If the current node exceeds this limit, recursion unwinds back to the alternative path
- As recursion unwinds, RBFS replaces the f value of each node along the path with the best f -value of its children
- RBFS retains the f -value of the best leaf in the forgotten subtree and can decide whether it should reexpand that subtree at a later time

Recursive Best-First Search (RBFS)

```
1.  Function RBFS(problem, node, f_limit)
2.      if (node.state == goal) return node
3.      successors = Expand(node, problem)
4.      if (successors.empty()) return (failure, infinity)
5.      for s in successors:
6.          s.f = max(s.g + h(s), node.f)
7.      repeat:
8.          best_node = min_f_node(successors)
9.          if (best_node.f > f_limit) return (failure, best.f)
10.         alt_node = second_lowest_f(successors)
11.         result, best_node.f = RBFS(problem, best_node, min(f_limit, alt_node))
12.         if (result != failure) return result
```

RBFS Performance

- Like A^* , RBFS is optimal if the heuristic function $h(n)$ is admissible
- Memory usage is linear in search depth $O(d)$
- Time complexity is difficult to analyze
 - Accuracy of heuristics?
 - How often does the best path change?

Memory-Bounded A^*

- Simple Memory-Bounded A^* (SMA*)
- Use all available system memory
- When all memory used and we need to add a node, erase the highest f-value node
- SMA* then backs up the value of the forgotten node to its parent
- The ancestor of a forgotten subtree knows the quality of the best path in that subtree
- In case of ties, SMA* **selects the newest** node generated, and **deletes the oldest** node generated
- SMA* is complete/optimal only if depth of shallowest goal < memory

Branch and Bound

Branch & Bound Algorithms

- “Any-time” search algorithm
 - Will continue to find increasingly better solutions until the you stop the algorithm
 - If run to completion, will find optimal solution
- Choose an initial search cutoff bound and search all solutions up to that bound, when a solution is found, update the bound to the cost of that solution

Depth-First Branch and Bound (DFBB)

```
best_path = []
```

```
bound = infinity
```

```
Function DFBB(problem, node, bound)
```

```
    if (node.state == goal)
```

```
        best_path = reconstruct_path(node)
```

```
        bound = path_cost(best_path)
```

```
    children = Expand(node, problem)
```

```
    sort(children, h()) # optional move-ordering
```

```
    for child in Expand(node, problem):
```

```
        if (child.g + h(child) > bound) continue # h must be admissible
```

```
        DFBB(problem, child, bound)
```

DFBB Example: Starcraft

Real-Time Strategy

- War-like Simulation
- Single / Multiplayer Games
- Most RTS Games:
 - Gather Resources
 - Build Town / Army
 - Combat With Enemies



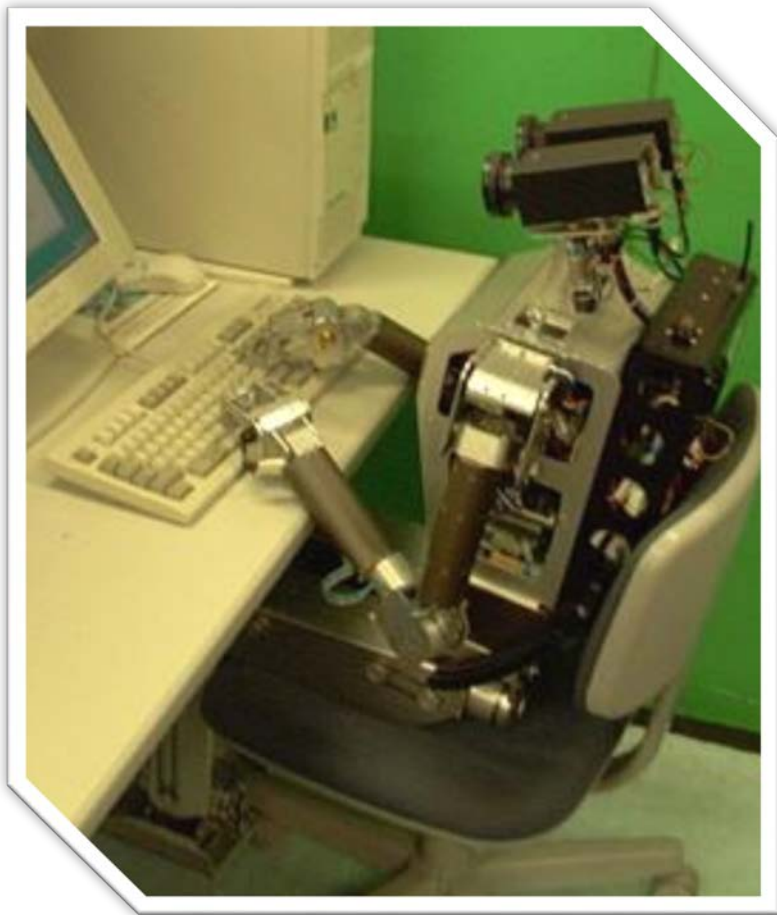
Properties of RTS

- Real-Time
- Simultaneous Move
- Non-Deterministic
- Imperfect Information
- Multi-Unit Control
- Unknown Game Engine
- Action / State Space



Benefits of RTS AI

- Better In-Game AI
 - More intelligent NPCs
 - Better single player
- Create Offline Tools
 - Game balancing
 - Reduce human testing
- Apply to any game

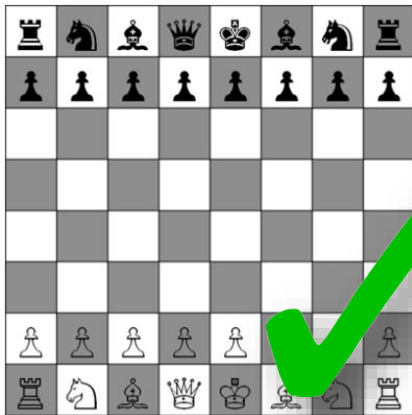


eSports

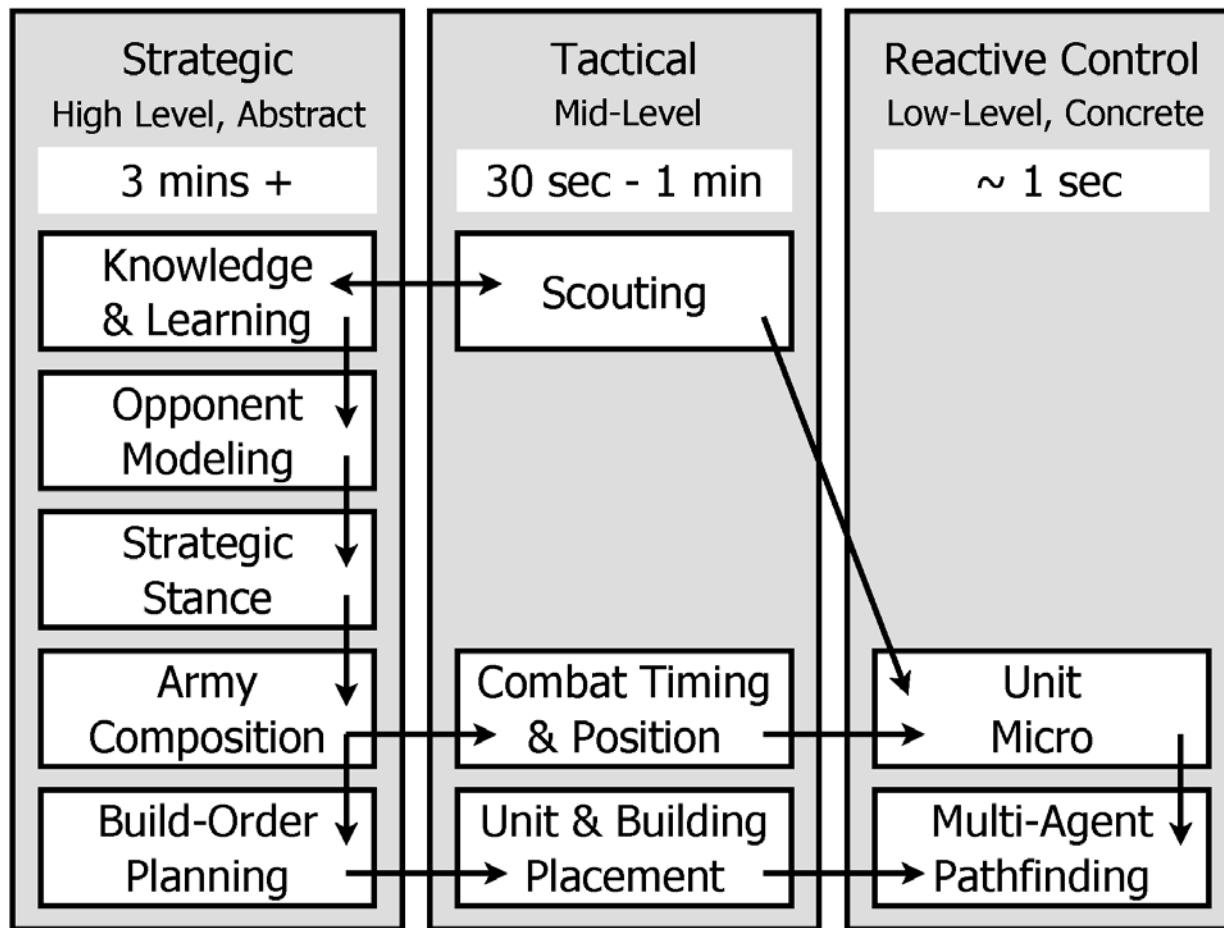


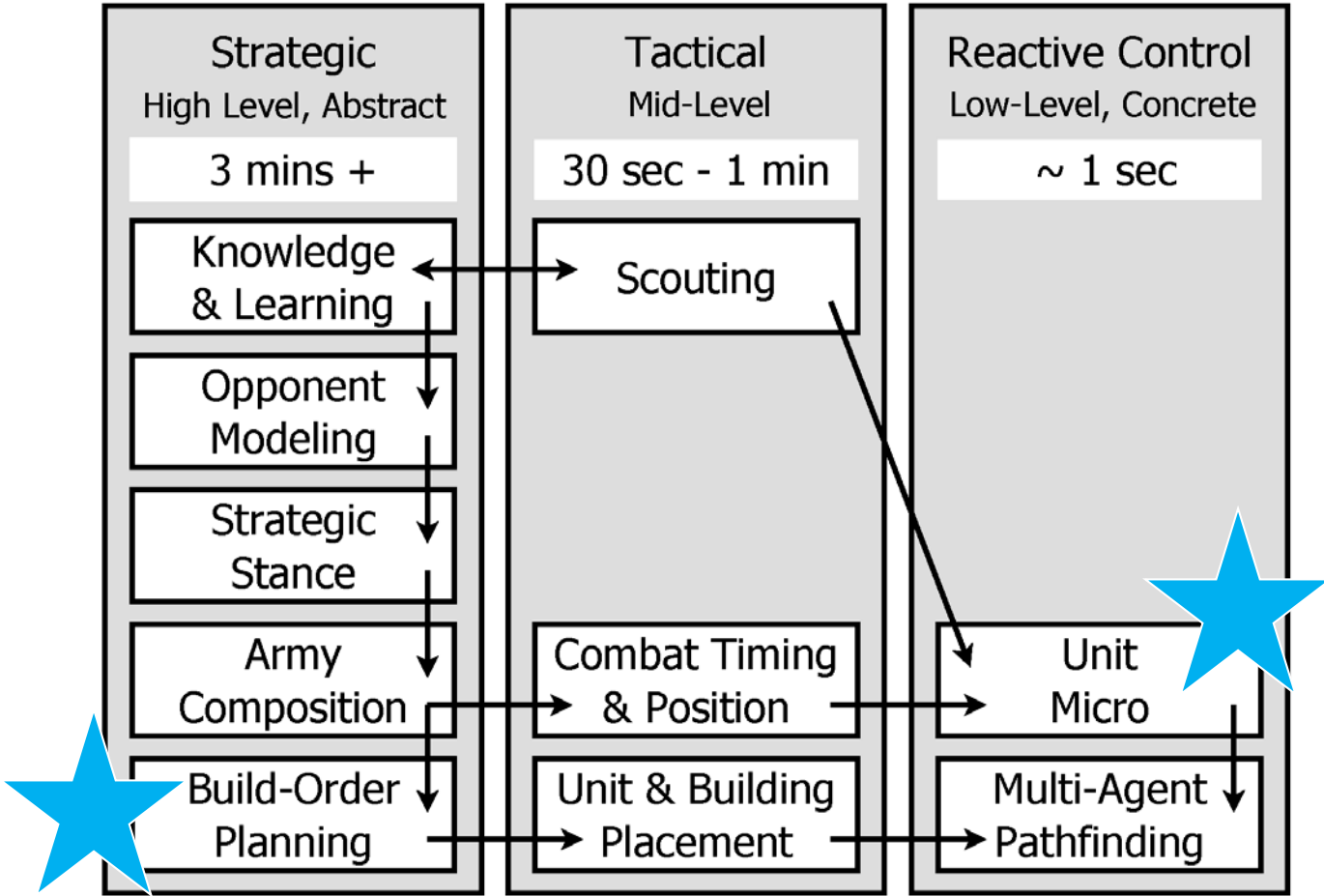


Human vs. Machine



StarCraft (Huge)



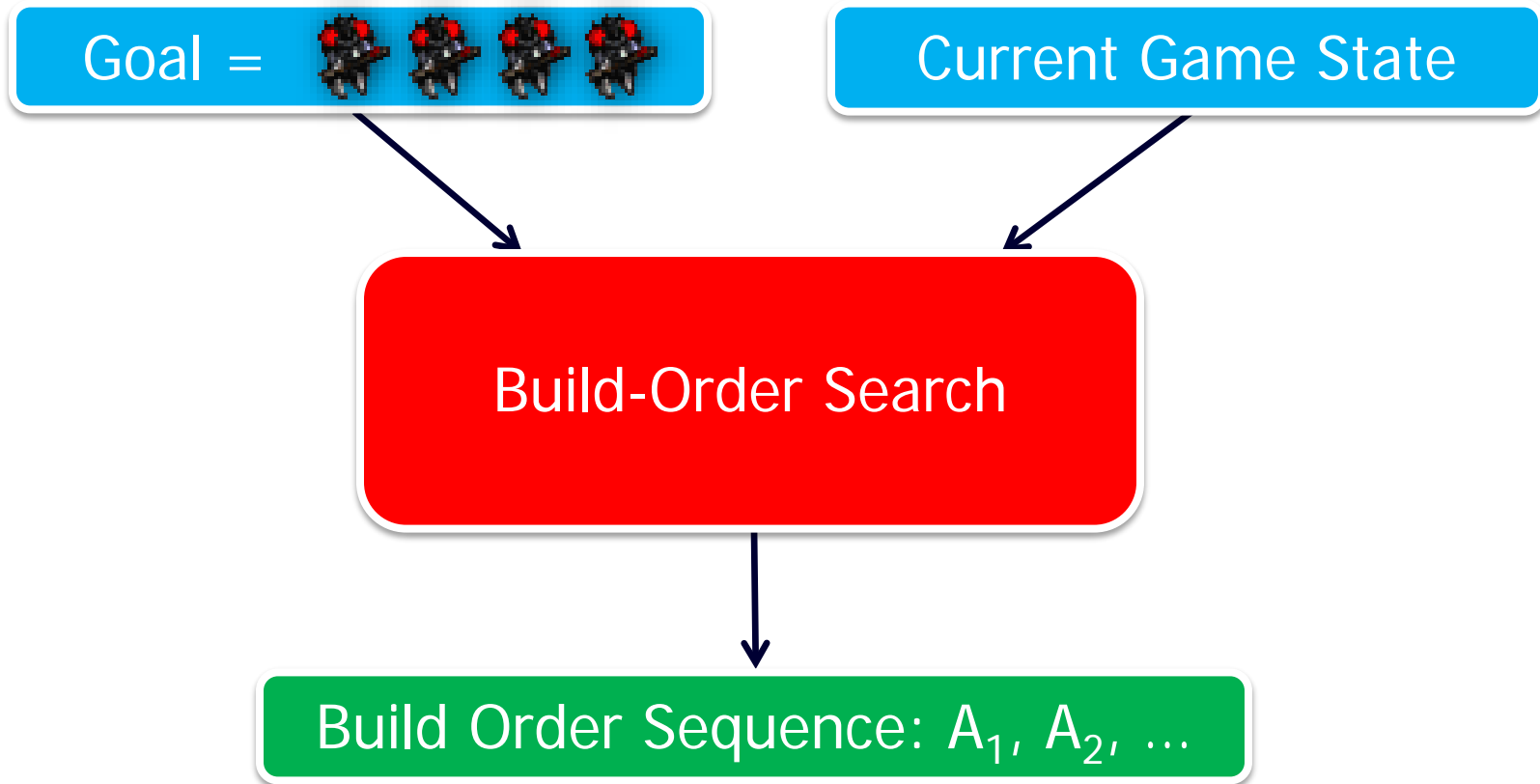


Build-Order Planning and Optimization

What is a Build-Order?

- Sequence of economic actions
- List of buildings / units to build in order
- Players memorize 'opening books'





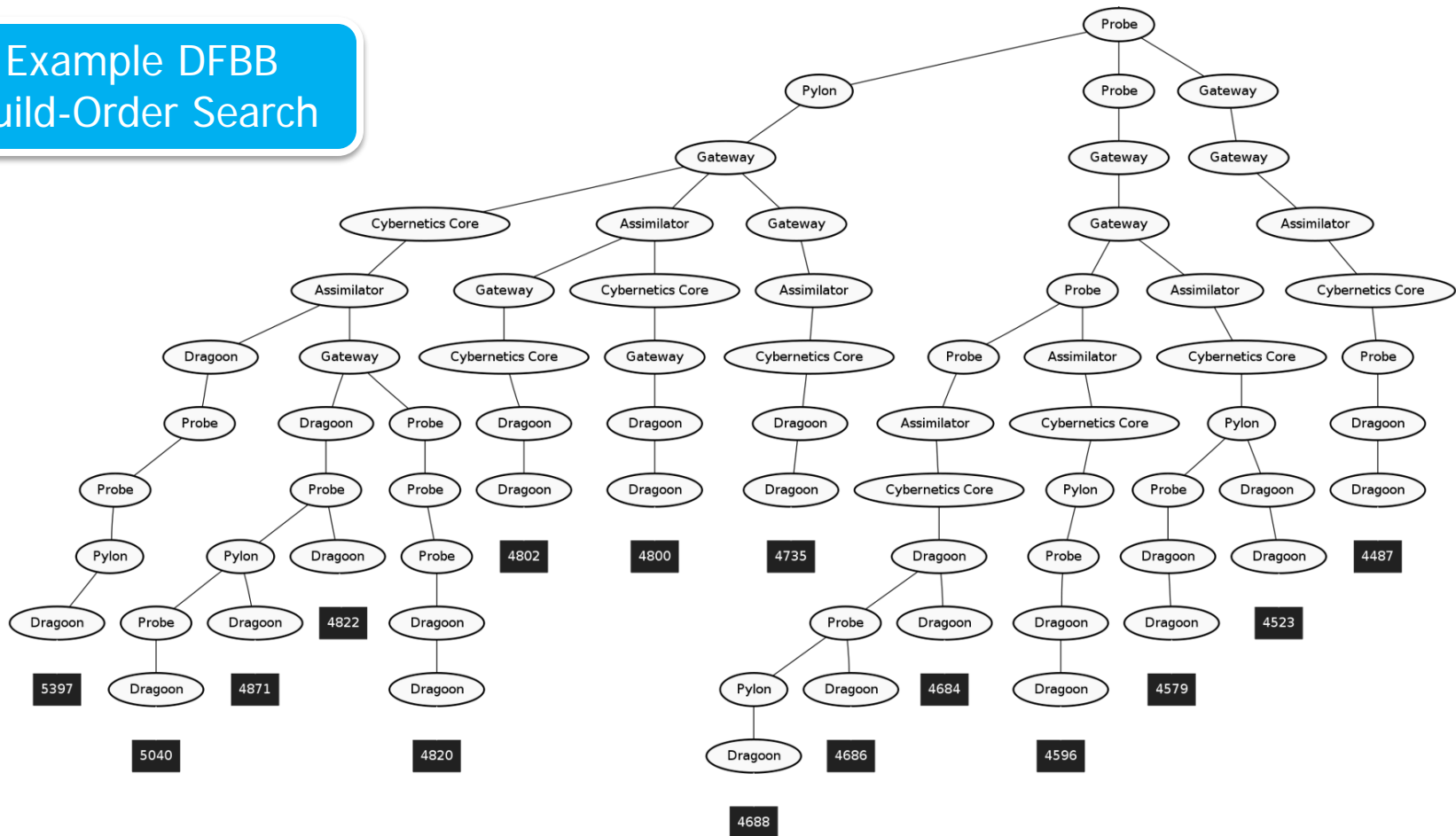
Algorithm - DFBB

- Depth-First Branch and Bound
- Optimize: Build-Order Makespan

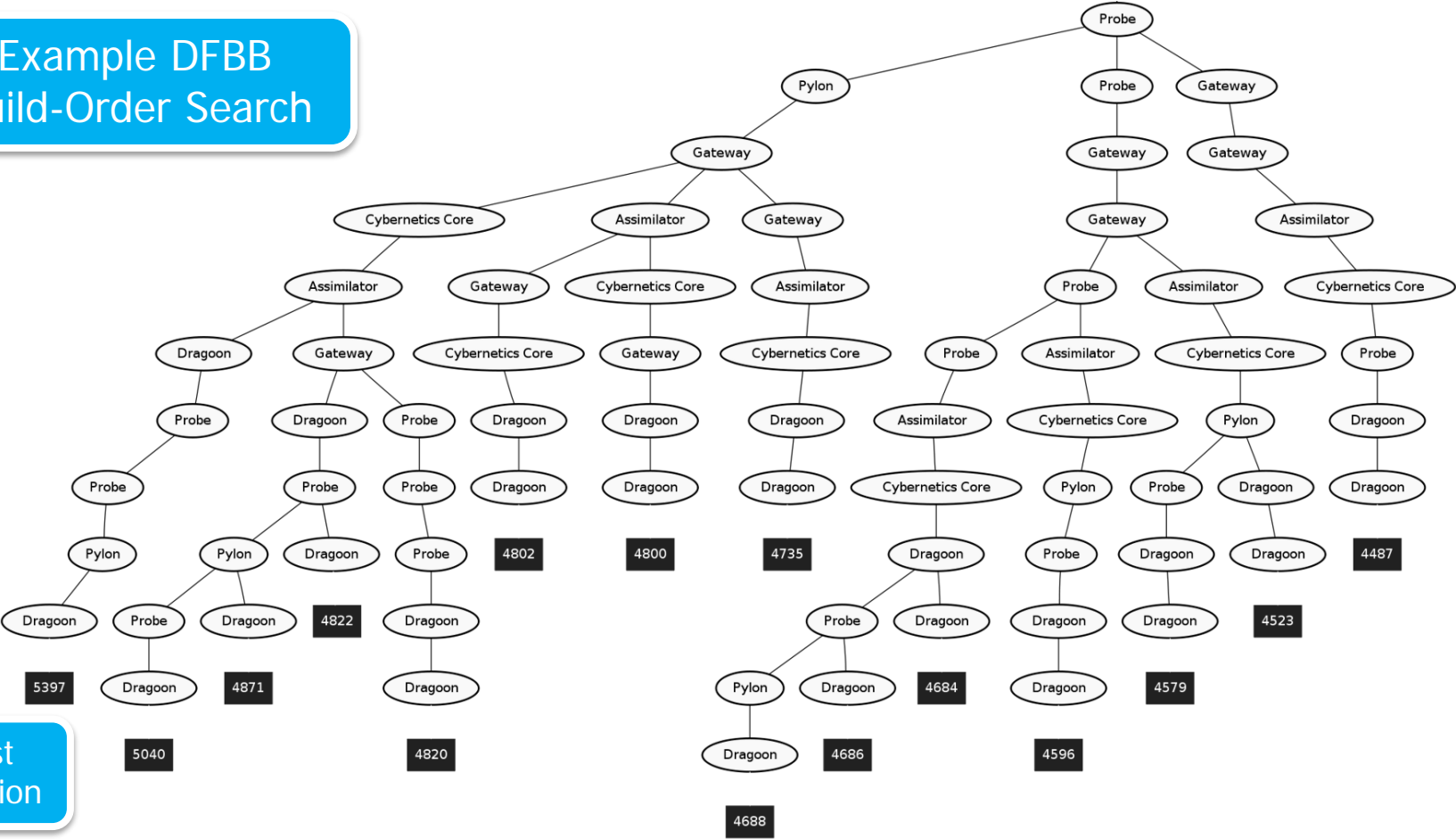
Algorithm - DFBB

- Depth-First Branch and Bound
- Optimize: Build-Order Makespan
- Low Memory Usage
- Any-Time
- Save / Load Search
- Upper + Lower Bound Heuristics

Example DFBB Build-Order Search

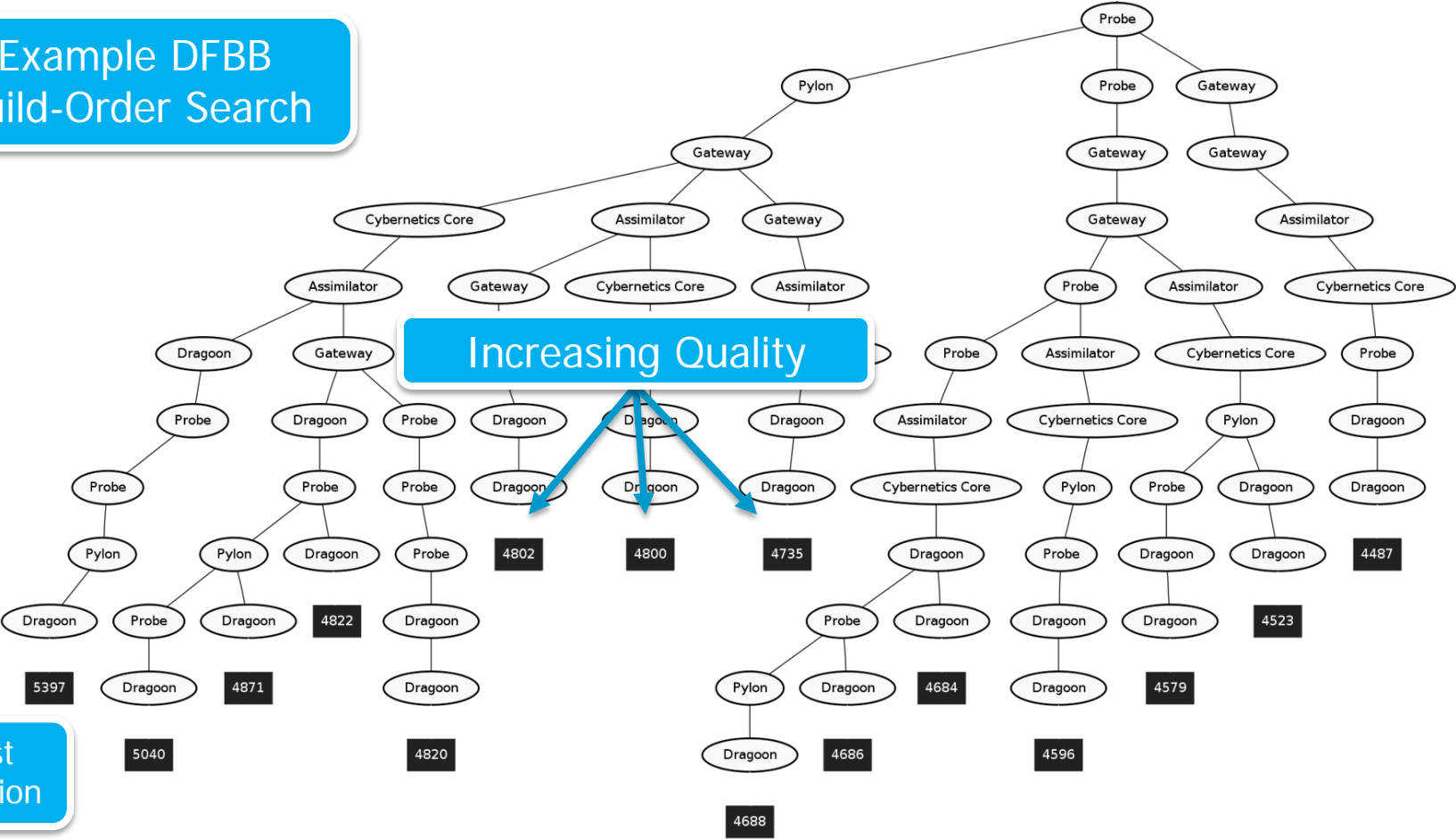


Example DFBB
Build-Order Search



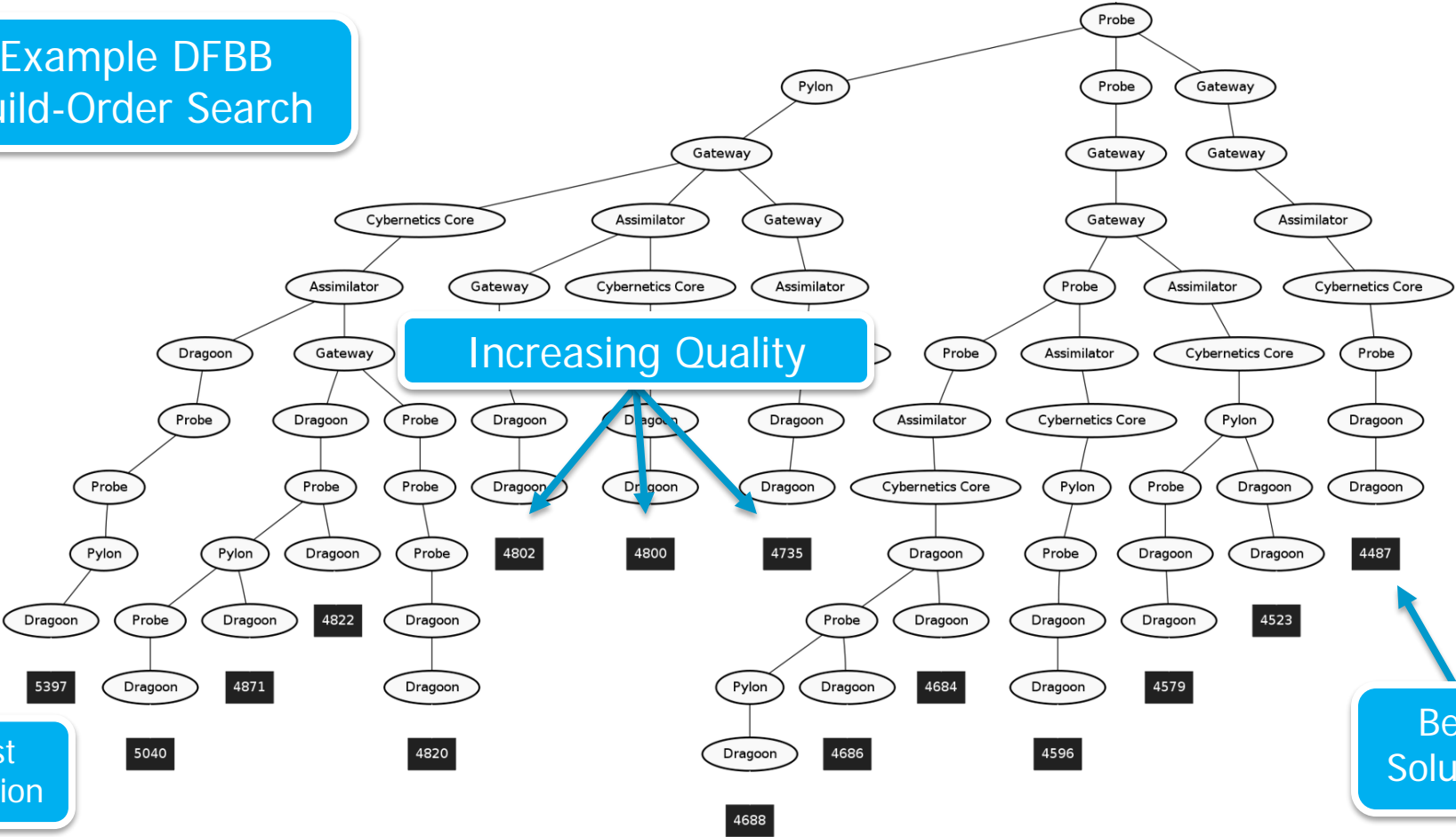
First
Solution

Example DFBB
Build-Order Search

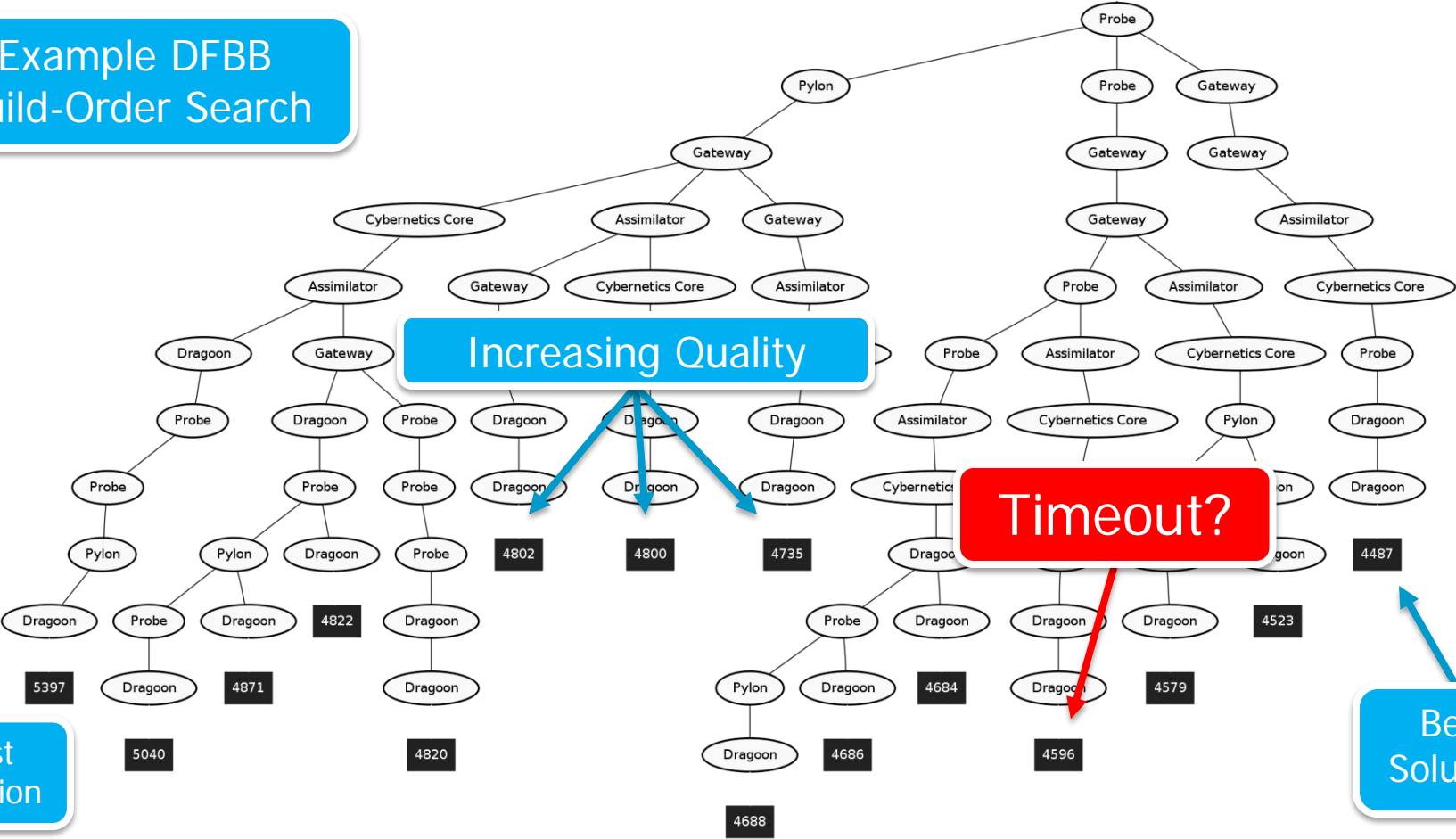


First
Solution

Example DFBB
Build-Order Search



Example DFBB
Build-Order Search



Pro Human Results

- Extract Pro Build-Orders
- Re-plan with BOSS
- Compare to original
- Results:
 - 100% of build orders solved
 - Faster than humans
 - Solve Time: 4% of makespan

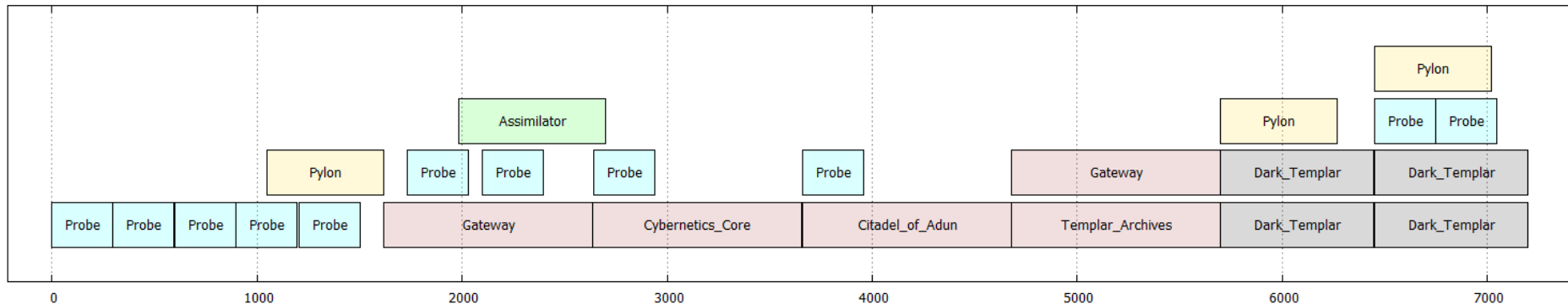


Build-Order Heuristics

- Resource Gathering Bound
- Landmark Heuristic
- Macro Actions
 - Repeat actions common in pro build orders

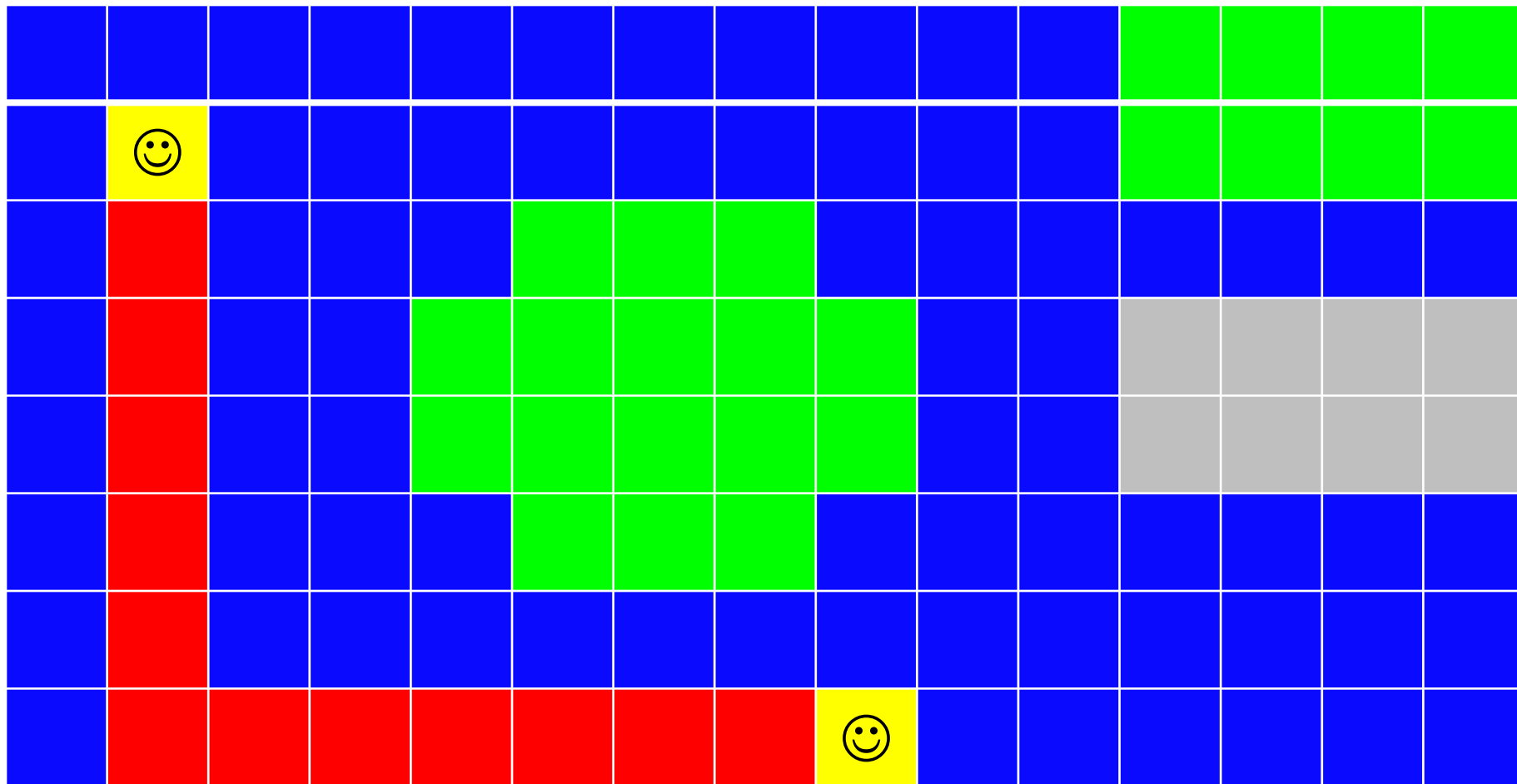
Landmark Heuristic

- How long do we need to build the sequential prerequisites of an action



Macro Actions

- Some problems have common sequences of actions that appear in optimal paths
- These commonly repeated action sequences are called “macro actions”
- Adding macro actions to the search can sometimes help us “shortcut” to better nodes further down the optimal path

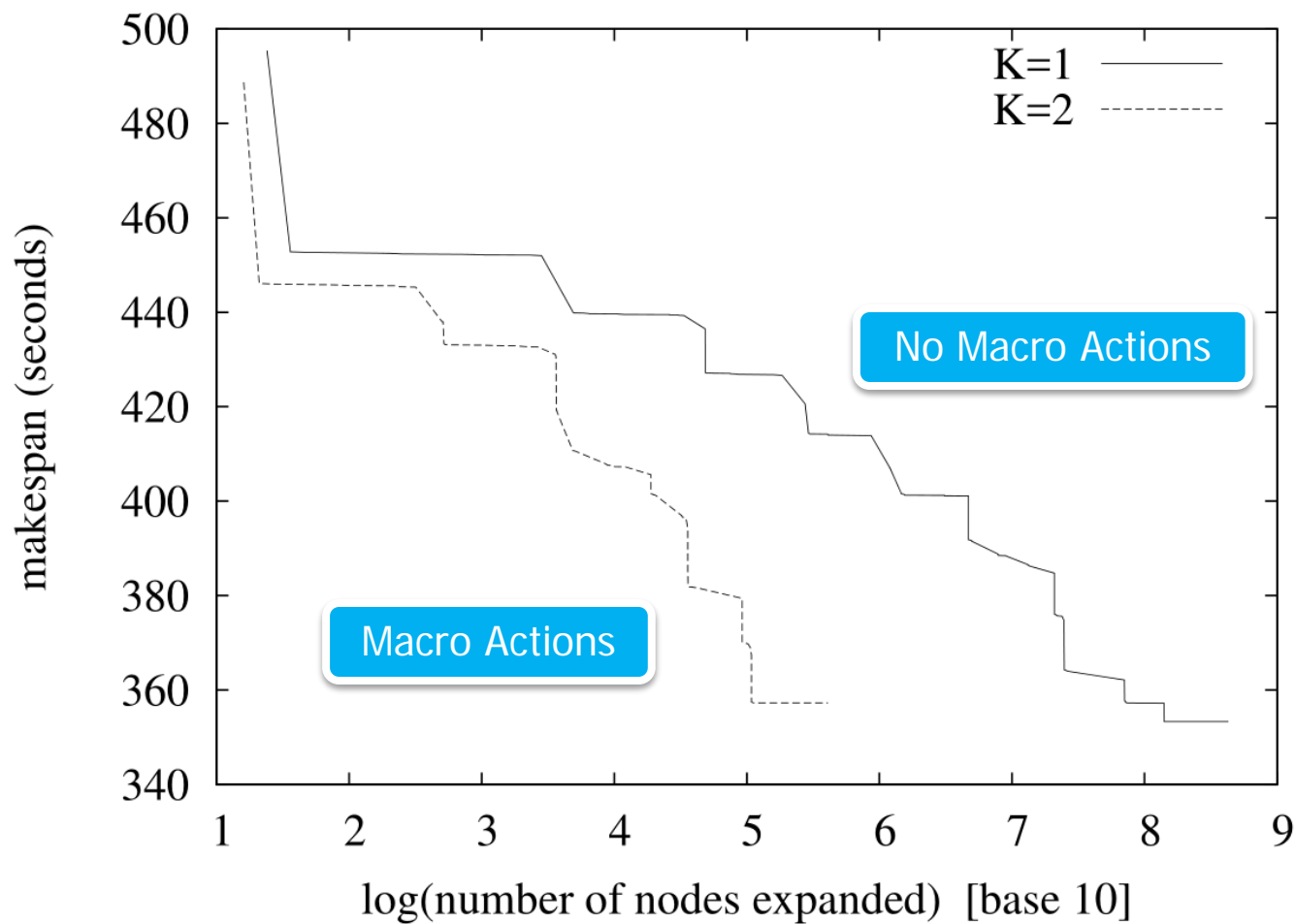


Macro Actions

- Macro actions are built of atomic actions
- They have cost equal to the sum of the actions contained within them
- Example: $> > ^$
 - Built From: Right Right Up
 - Cost: $100 + 100 + 141 = 341$

Macro Actions

- Trade-off: branching factor vs. depth
- We must keep all atomic actions to ensure that the optimal solution is still found
- Macro actions search deeper into the tree with a single transition, but increase the branching factor of a given node
- They may speed up the search, but a poor choice of macro action can also slow down the search



Build-Order Planning Impact

- Implemented into UAlbertaBot
 - Used in other bots in past few years
- All build-orders planned in real-time
 - First real-time planning solution
- Ran in real time with no time-outs
- UAlbertaBot placed 2nd when implemented

StarCraft Build-Order Visualization

Help

Select Race

Protoss

Protoss_Probe

Add to Build Order

Add to Initial State

Protoss_Probe

Producer: Protoss_Nexus

Minerals: 50

Build Time: 300

Gas: 0

HP / Shield: 20 / 20

Build-Order

View Details

Protoss_Probe
 Protoss_Probe
 Protoss_Probe
 Protoss_Probe
 Protoss_Pylon
 Protoss_Probe
 Protoss_Probe
 Protoss_Probe
 Protoss_Gateway

Clear Selected

Clear All

Load Build-Order

Save Build-Order

Initial State

View Details

Protoss_Probe
 Protoss_Probe
 Protoss_Probe
 Protoss_Probe
 Protoss_Nexus

Clear Selected

Clear All

Load State

Save State

Starting Minerals

50

Starting Gas

0

View Final State

Generate Graphs

Army Value Graph

Resource Graphs

Build-Order Graph

