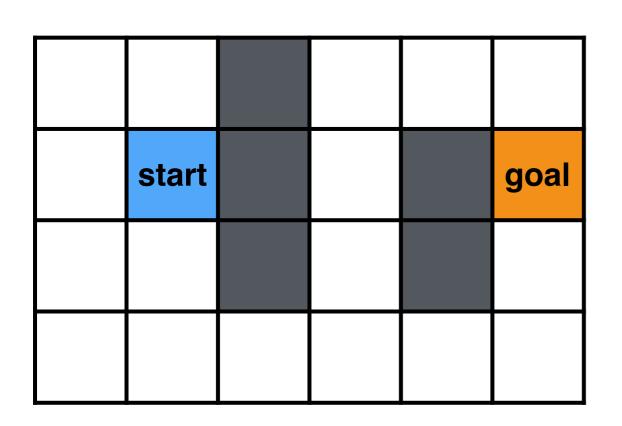
# Robot Autonomy

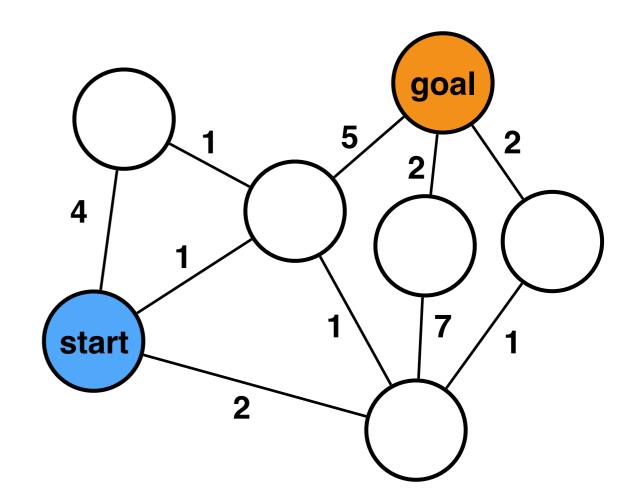
# Lecture 09: Discrete Search

Oliver Kroemer

#### Motivation

Discretized motion planning problems





- Edges may have uniform or different distances/costs
- Want to find a (short/cheap) path from start to goal state

#### Discrete Piano Mover's Problem

Discrete state

$$x \in X$$

Discrete actions

$$u \in U(x)$$

State transition function

$$f(x,u) = x'$$

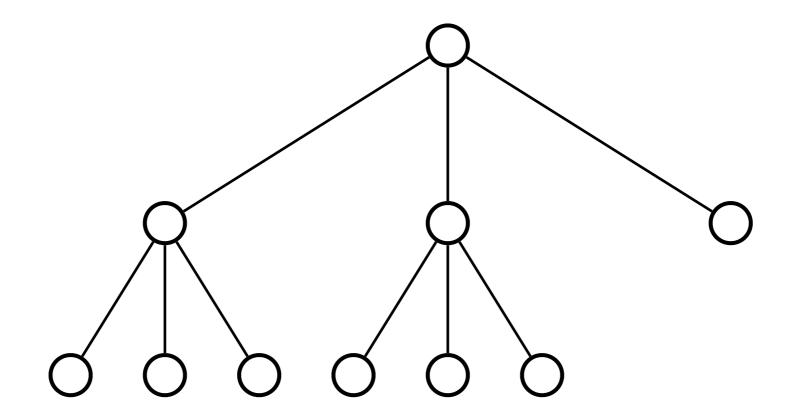
traverse edges of graph or move to neighbouring grid cell

## Forward Search with Priority Queue

```
Priority queue Q = \text{empty}
Q.insert(x_i) mark x_i as visited
while Q is not empty
       x \leftarrow Q.\text{GetFirst}()
       if x \in X_{qoal}
               return Success
       for all u \in U(x)
               x' \leftarrow f(x, u)
               if x' not visited
                       mark x' as visited
                       Q.insert(x')
```

return Failure

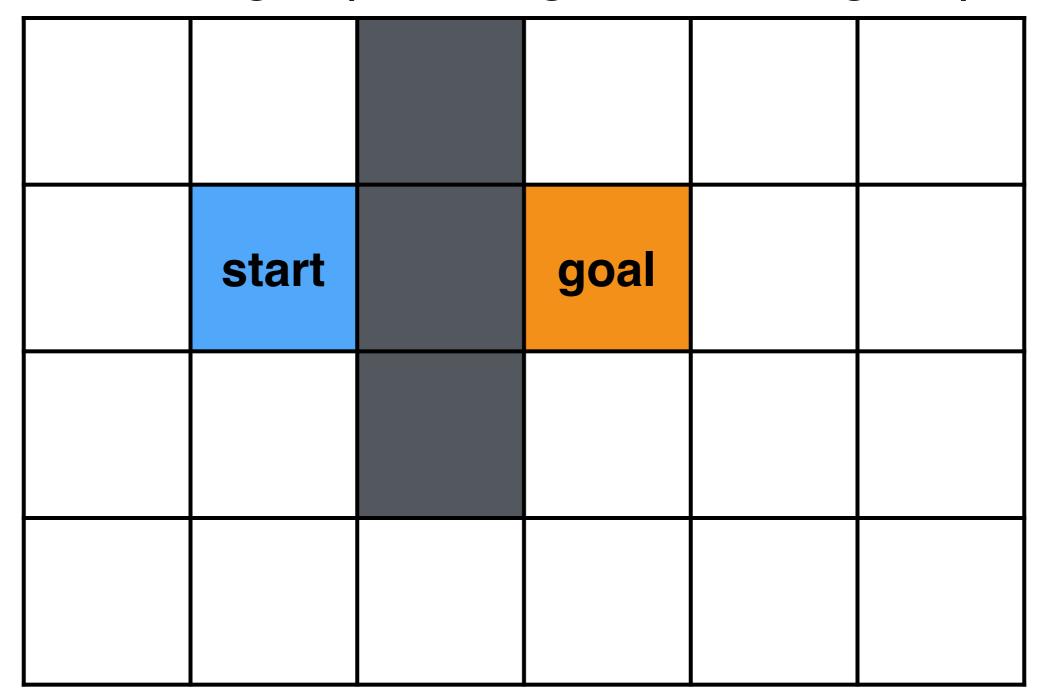
- Breadth-first search of states
- Expand states according to first-in-first-out queue



- Will find shortest path to goal
- Will expand to every state closer to the start

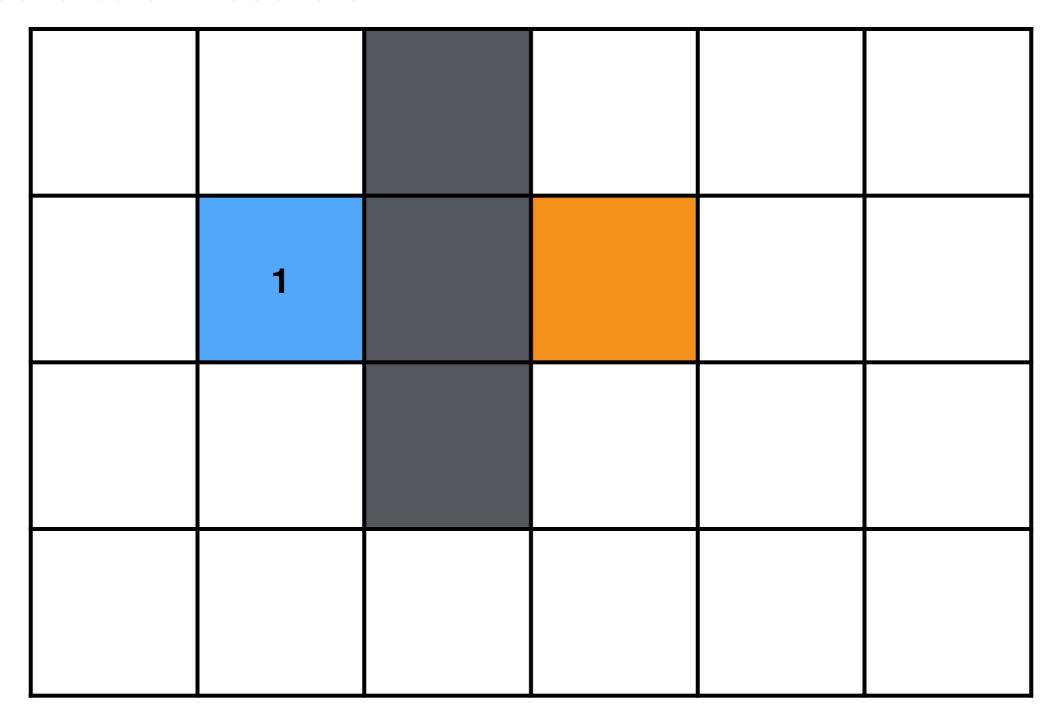
### Example Grid Problem

• Find start-to-goal path using down, left, right, up actions



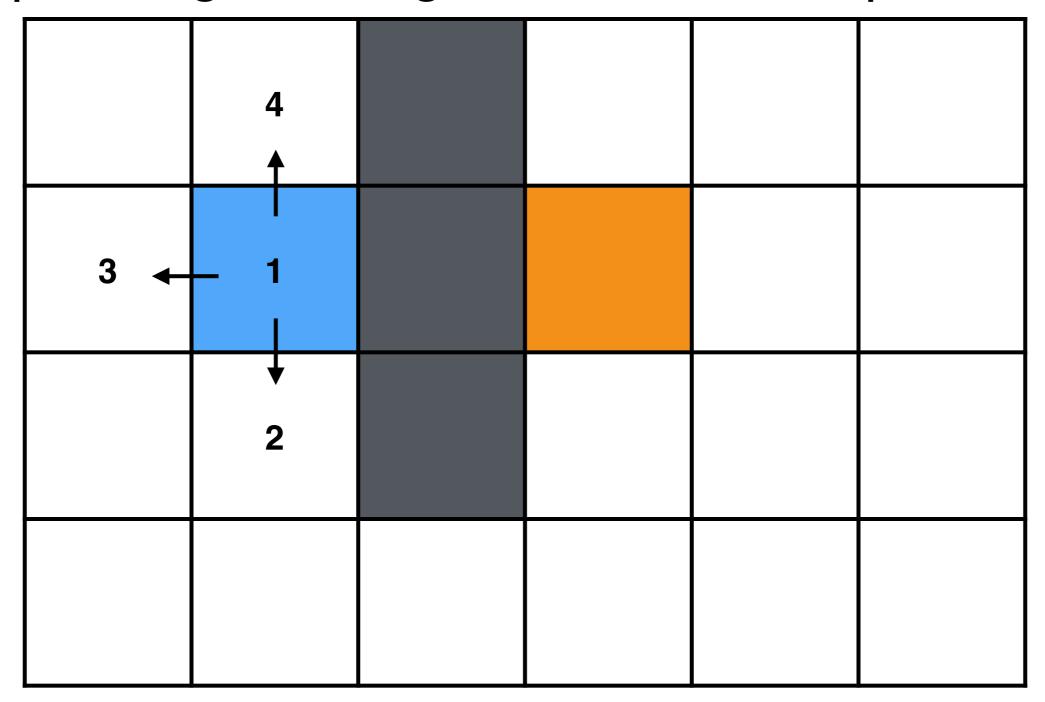
Compute obstacles online

Start at the start cell



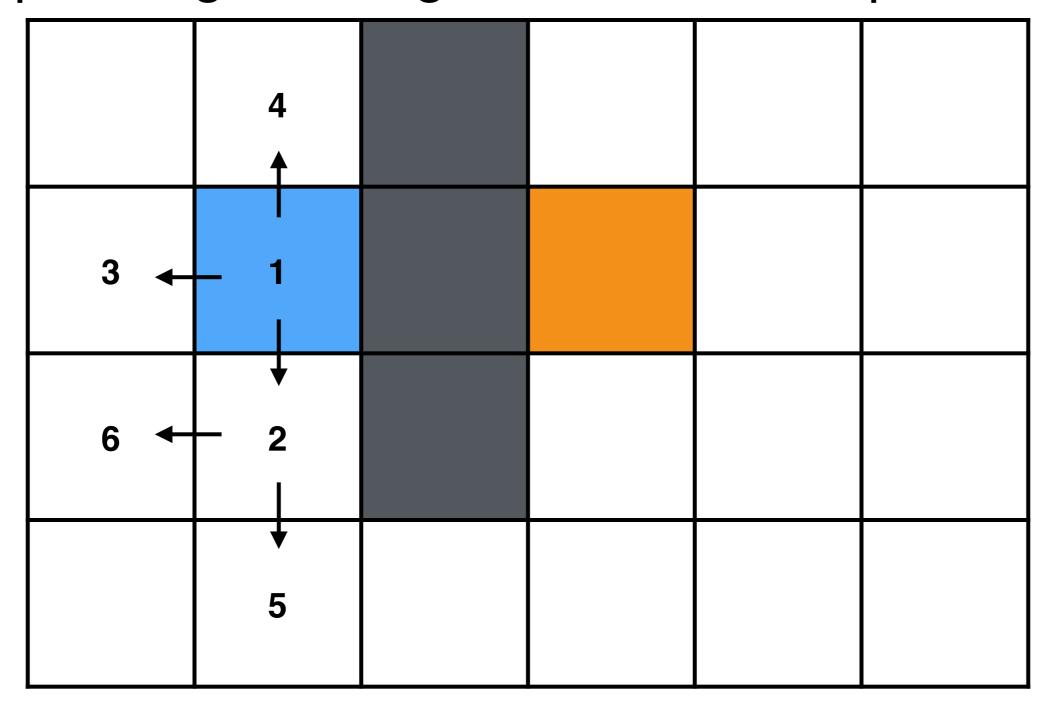
Q=[I]

• Expand neighbours right, down, left, and, up



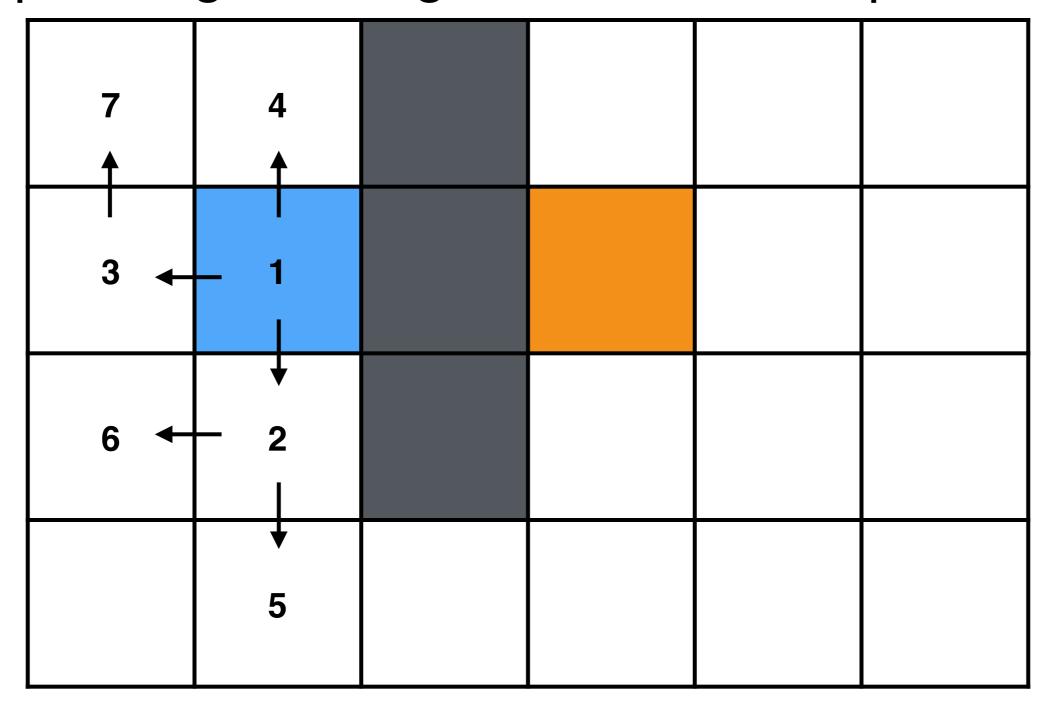
• Q=[1,2,3,4]

• Expand neighbours right, down, left, and, up



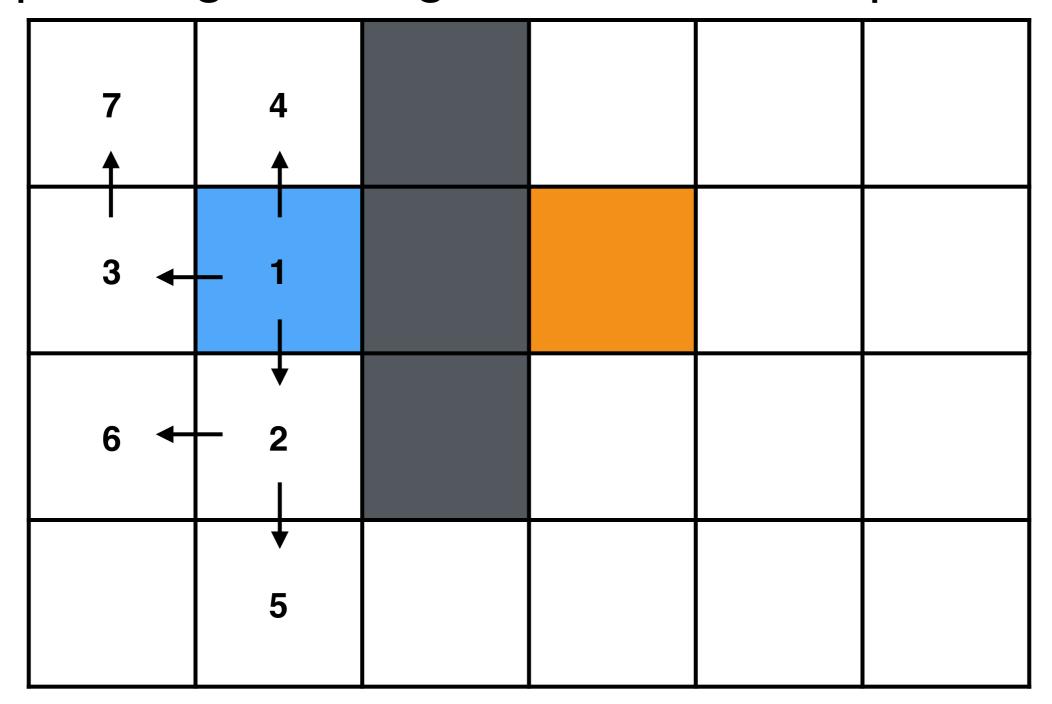
• Q=[2,3,4,5,6]

• Expand neighbours right, down, left, and, up



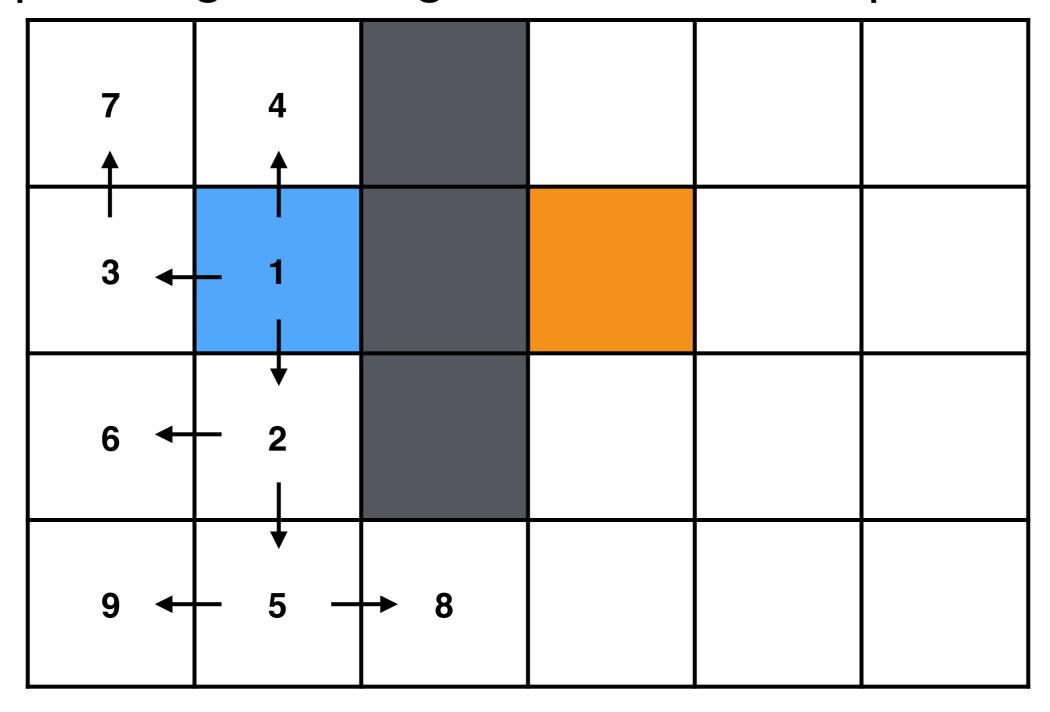
• Q=[3,4,5,6,7]

• Expand neighbours right, down, left, and, up



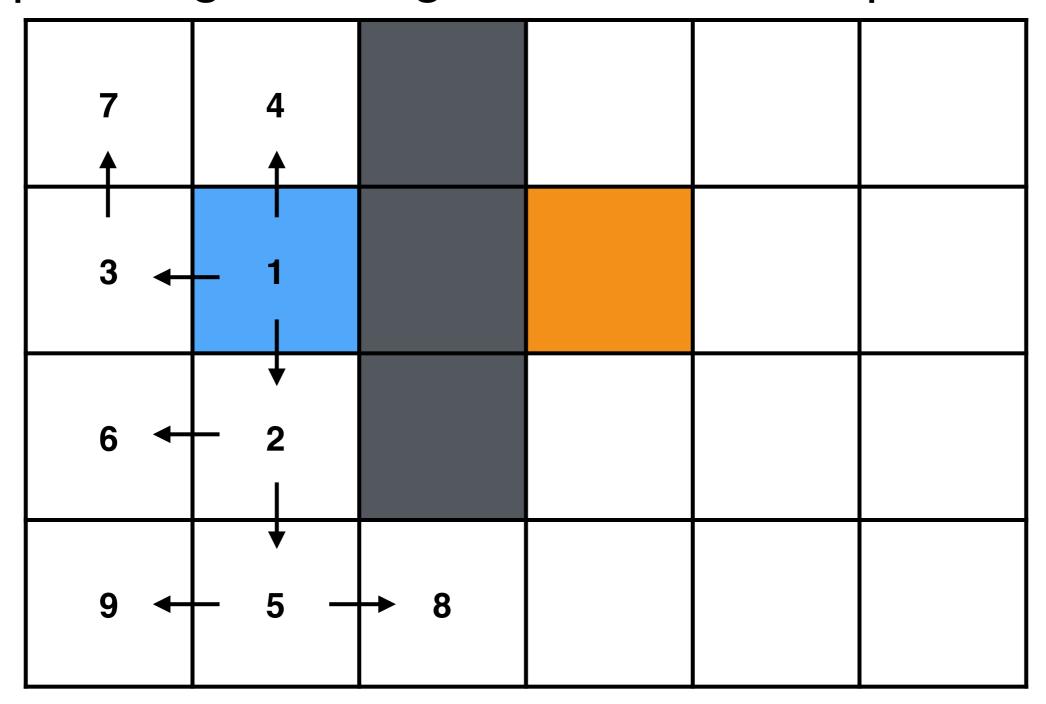
• Q=[4,5,6,7]

• Expand neighbours right, down, left, and, up



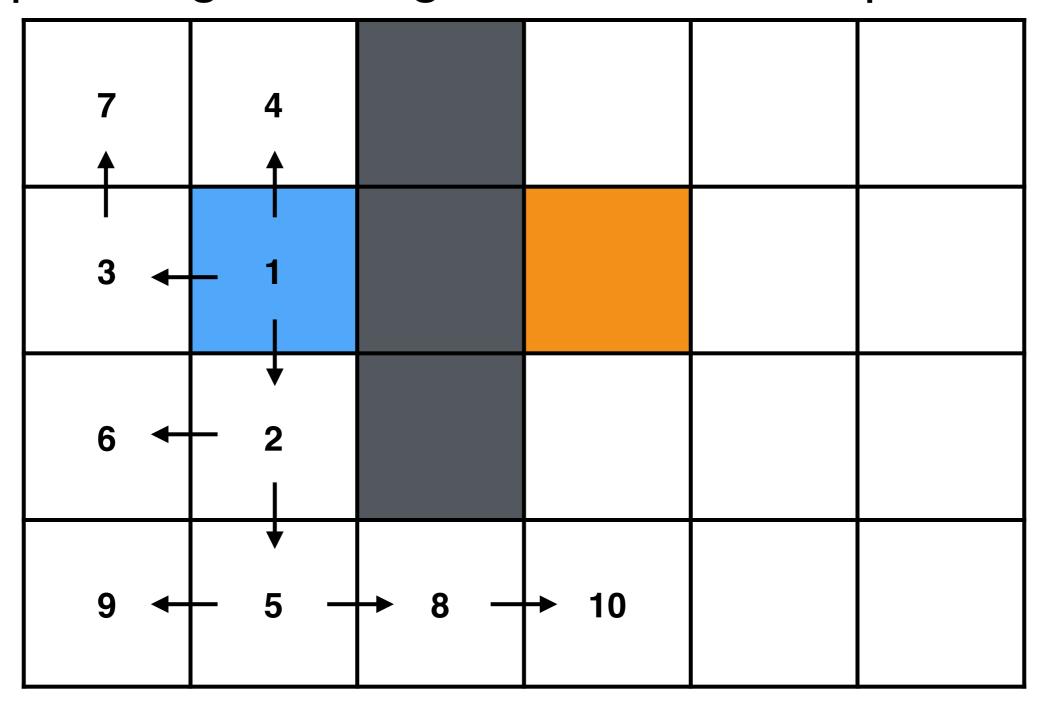
• Q=[5,6,7,8,9]

• Expand neighbours right, down, left, and, up



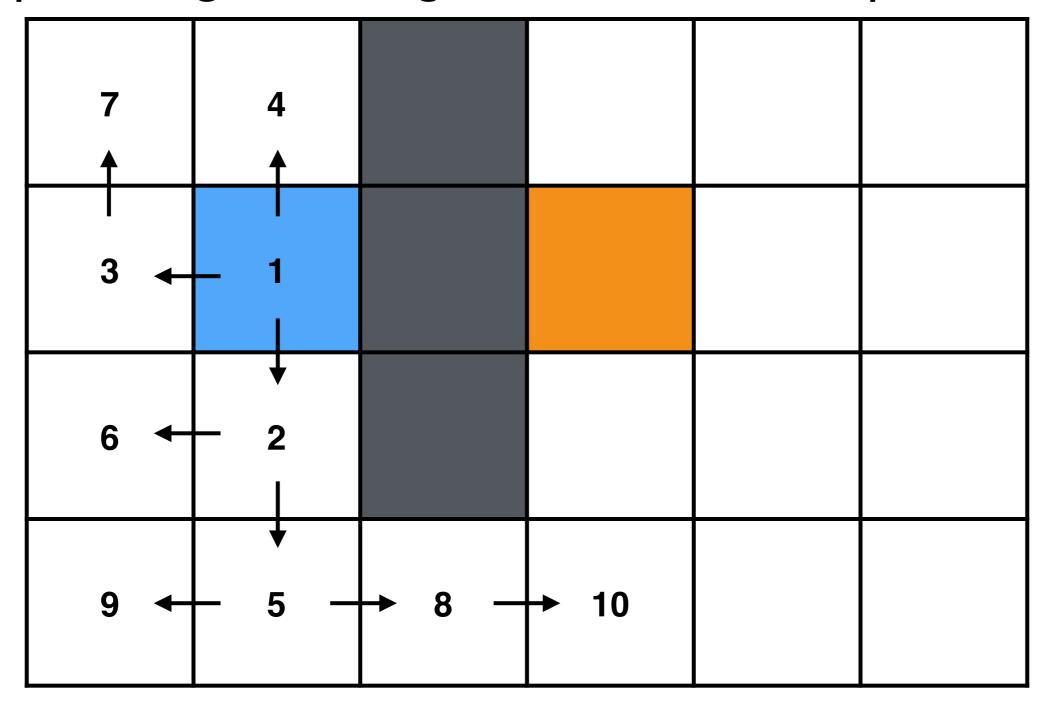
• Q=[6,7,8,9]

• Expand neighbours right, down, left, and, up



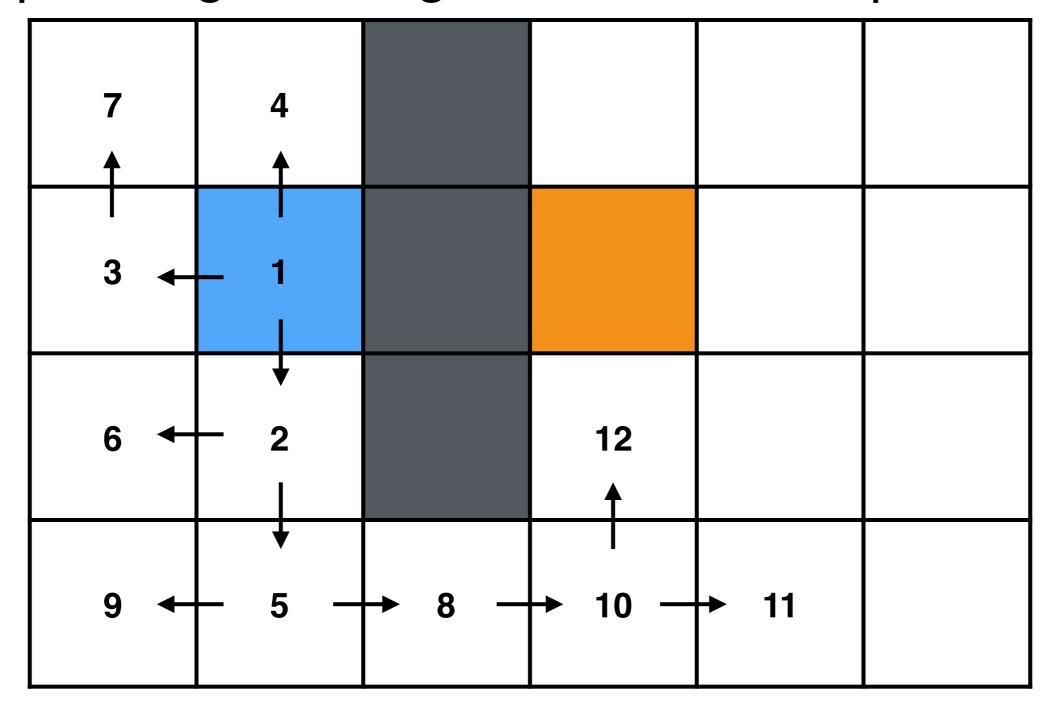
• Q=[8,9,10]

• Expand neighbours right, down, left, and, up



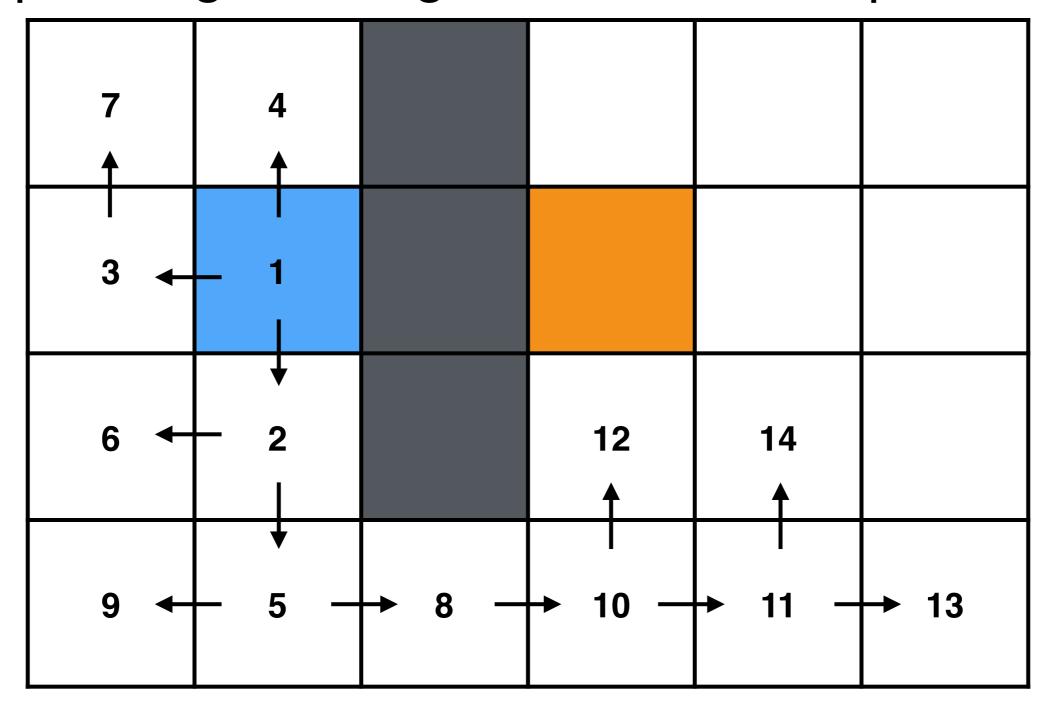
• Q=[9,10]

• Expand neighbours right, down, left, and, up



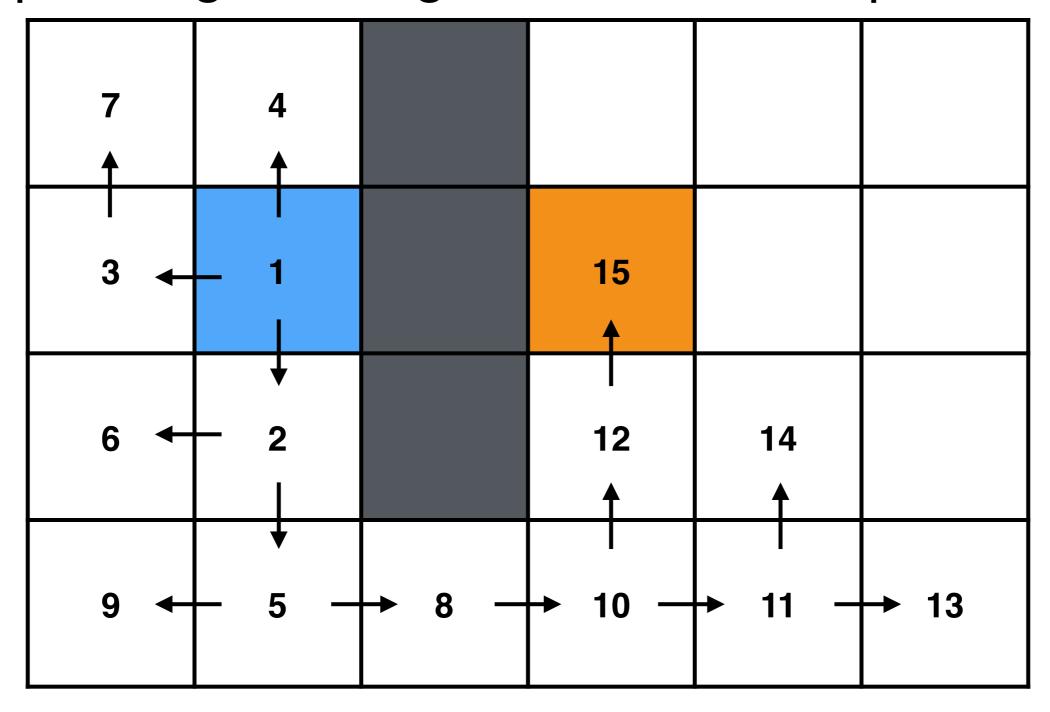
• Q=[10,11,12]

• Expand neighbours right, down, left, and, up



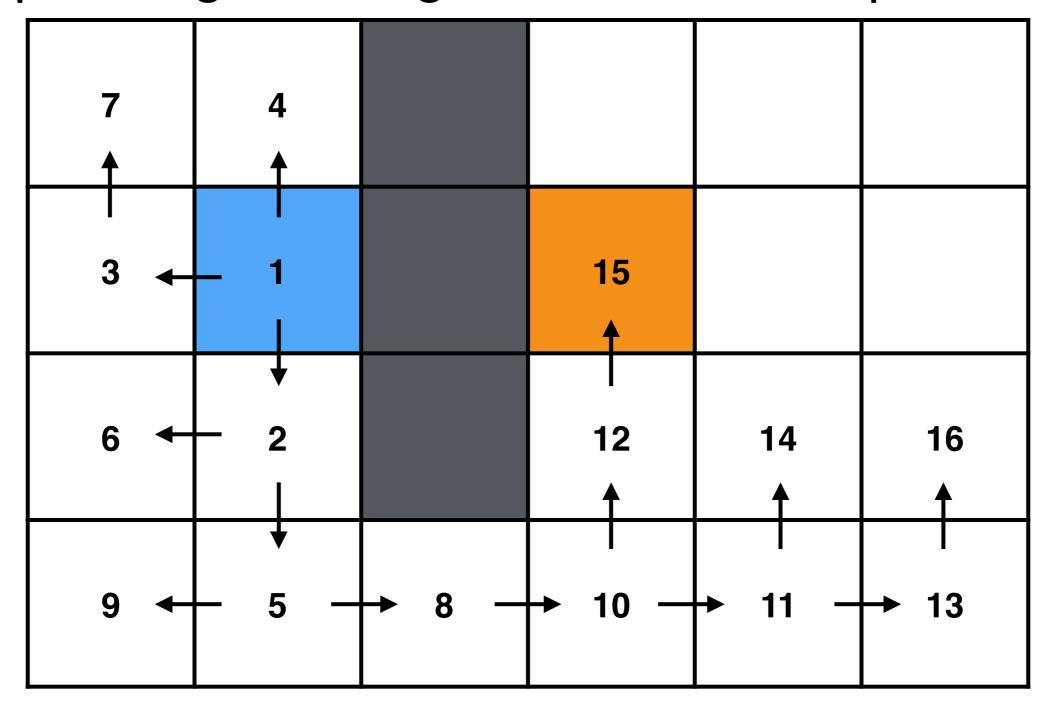
• Q=[11,12,13,14]

• Expand neighbours right, down, left, and, up



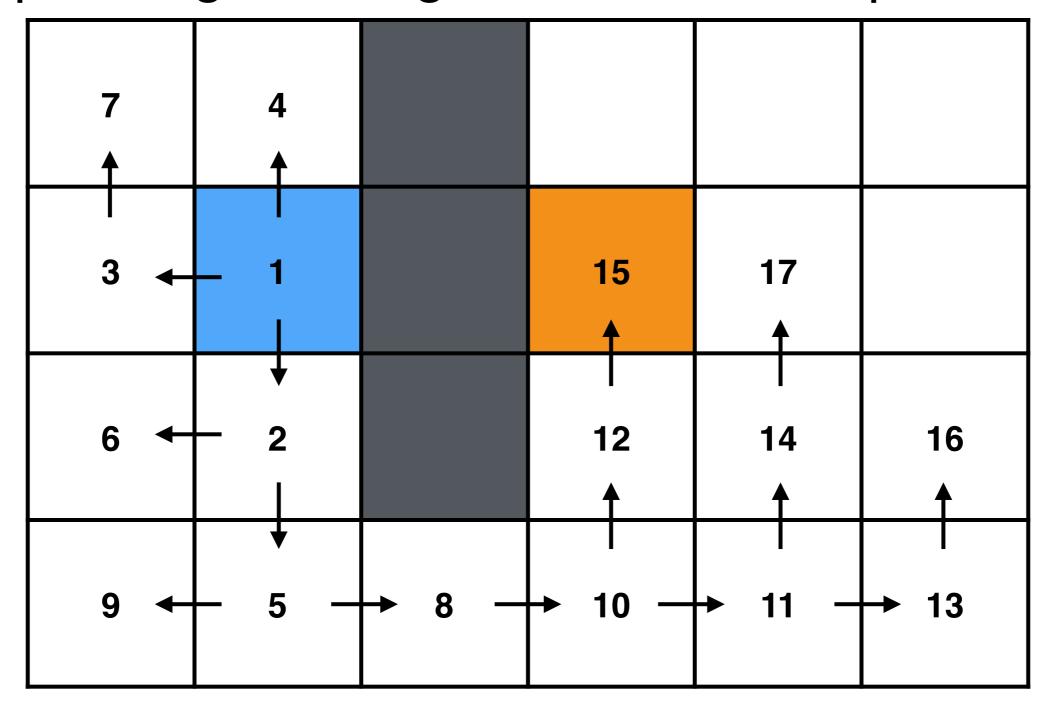
• Q=[12,13,14,15]

Expand neighbours right, down, left, and, up



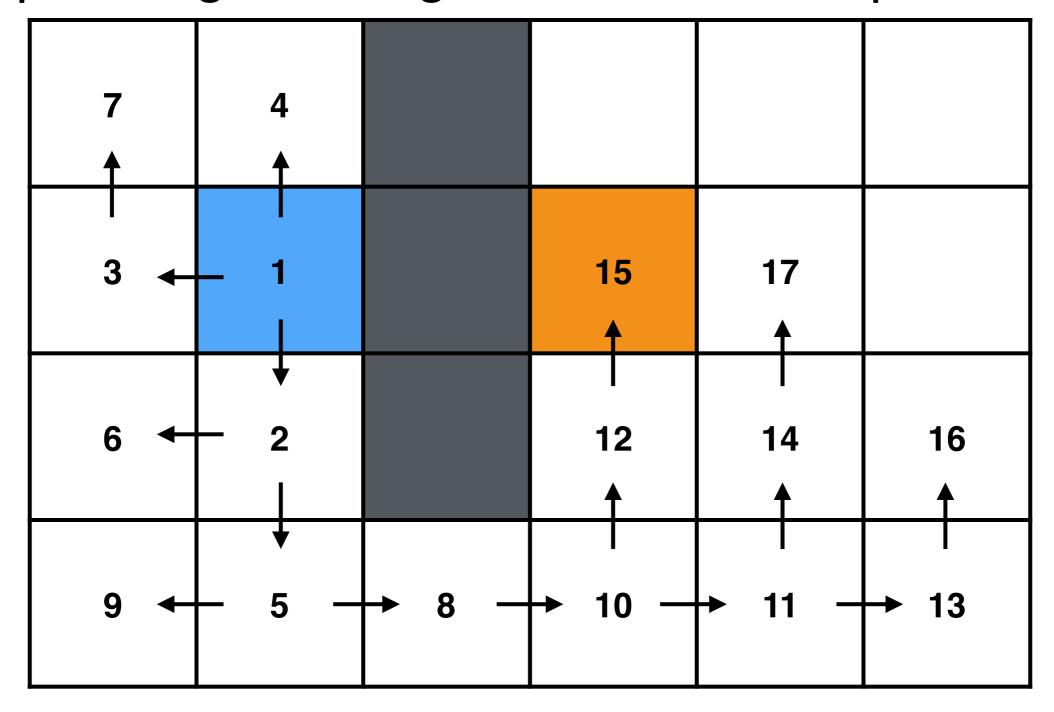
• Q=[13,14,15,16]

• Expand neighbours right, down, left, and, up



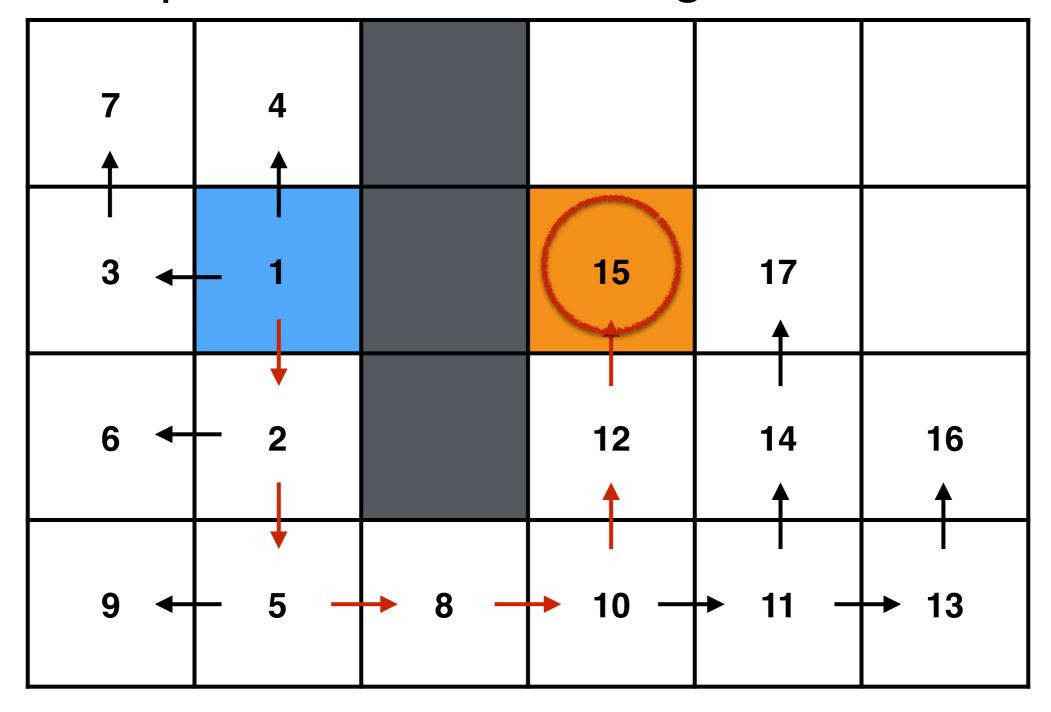
• Q=[14,15,16]

Expand neighbours right, down, left, and, up



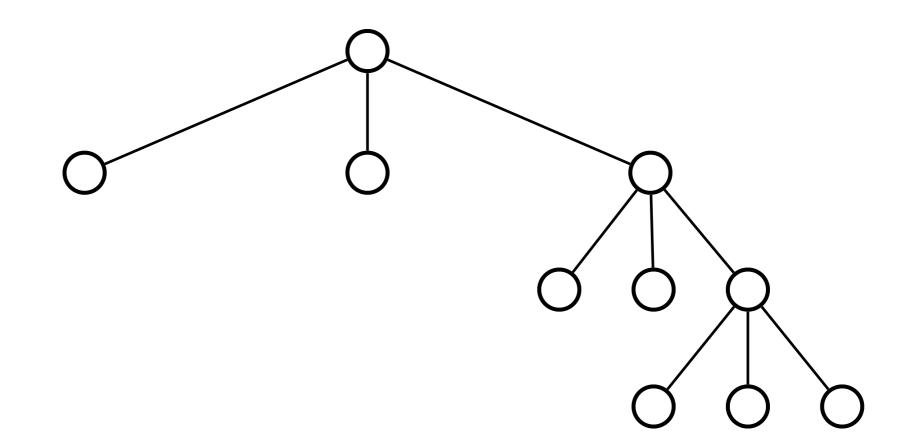
• Q=[15,16] GOAL!!!

Shortest path between start and goal



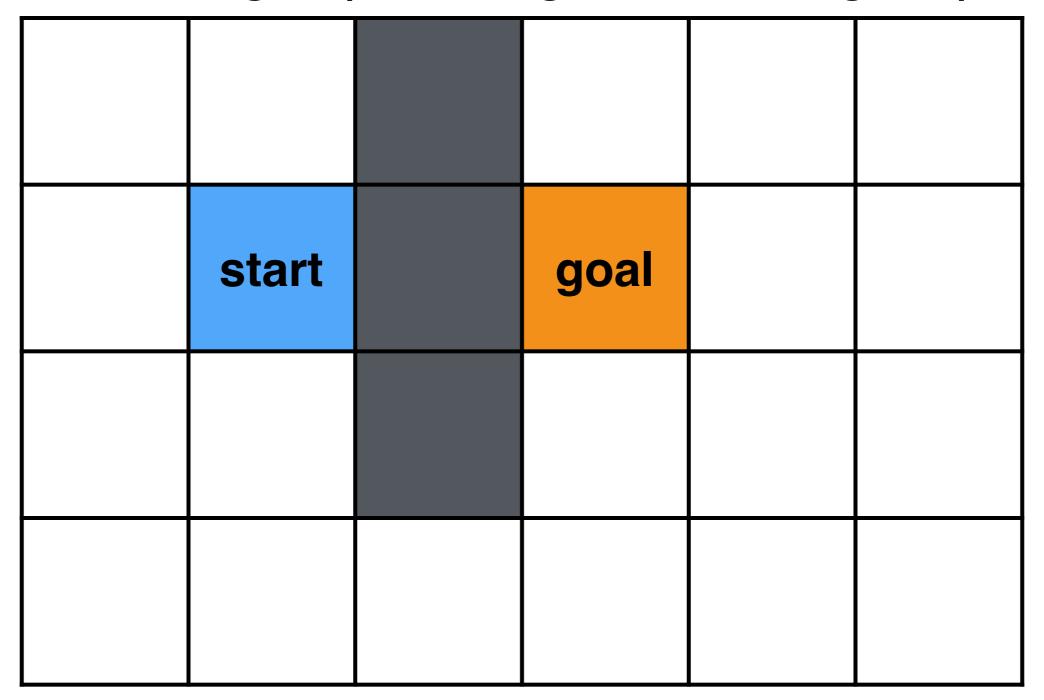
lowest value neighbour: lower numbers closer or equal

- Depth-first search of states
- Expand states according to last-in-first out queue



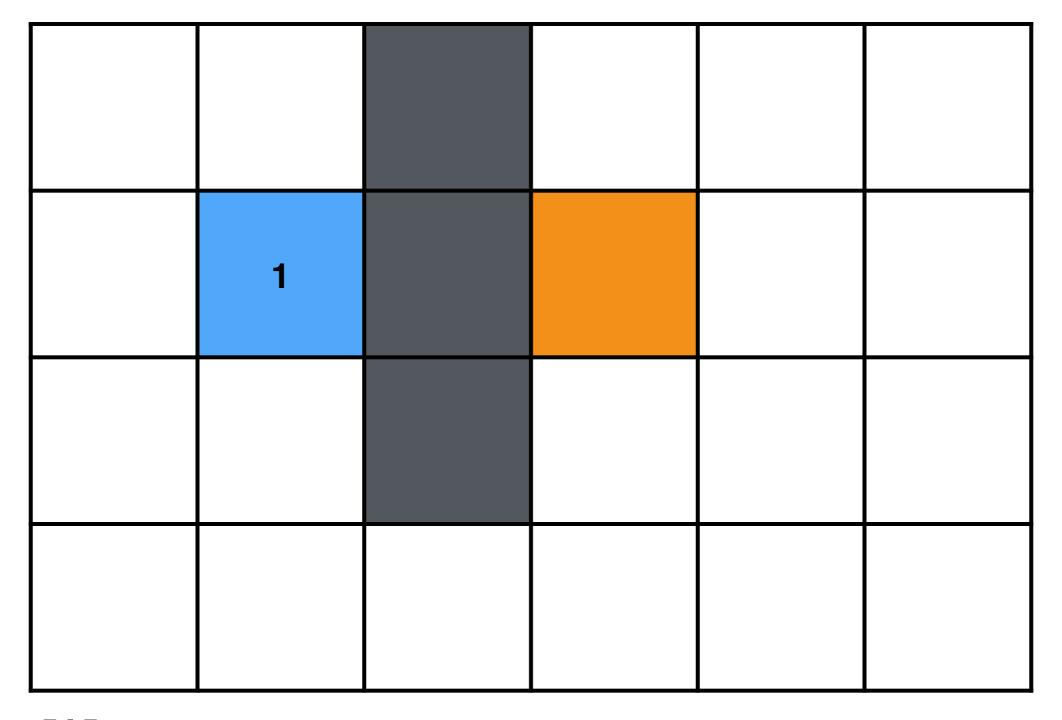
- Expand until a deadend or fixed depth (depth-limited)
- Quickly expands along one branch

• Find start-to-goal path using down, left, right, up actions



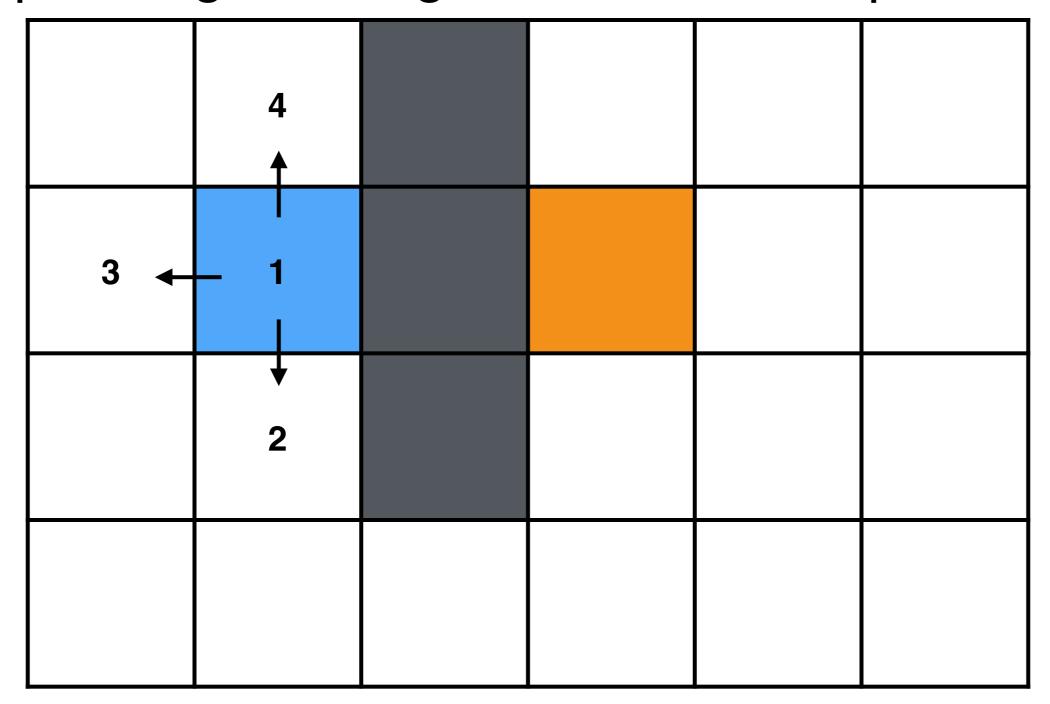
Compute X\_obs online

• Start at the start cell



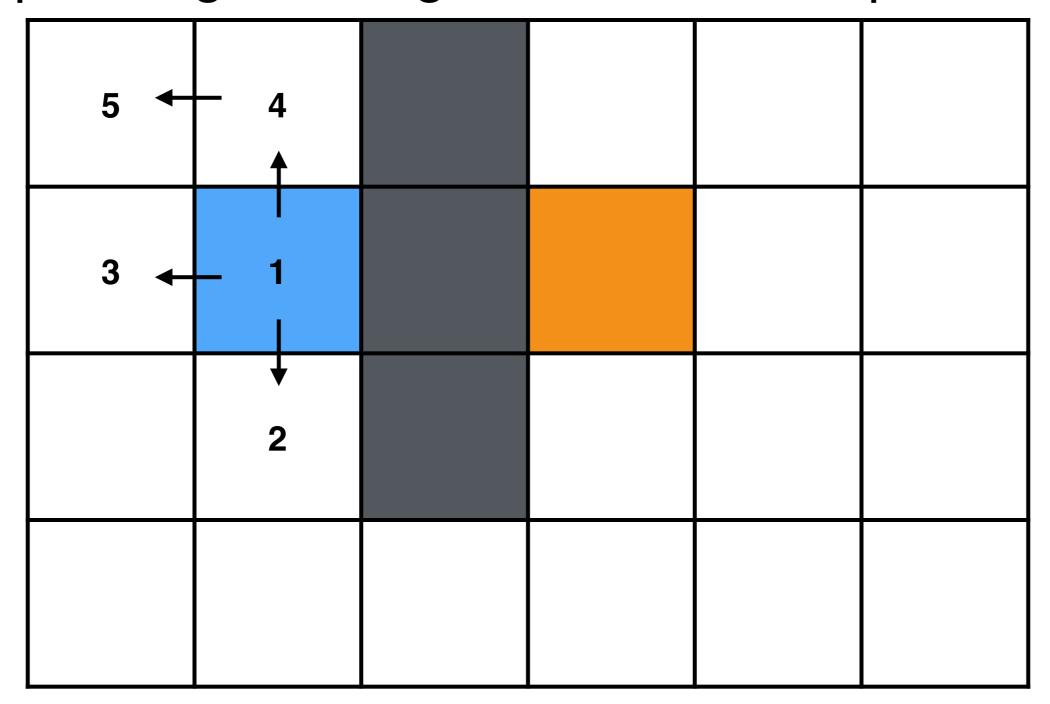
Q=[1]

• Expand neighbours right, down, left, and, up



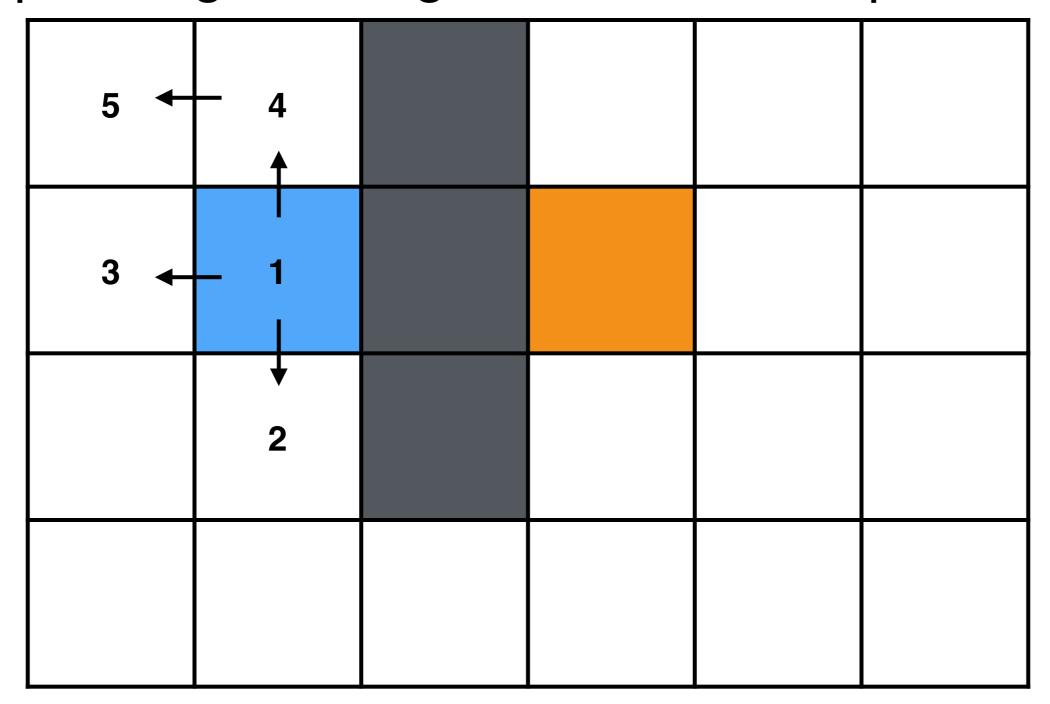
• Q=[1,2,3,4]

• Expand neighbours right, down, left, and, up



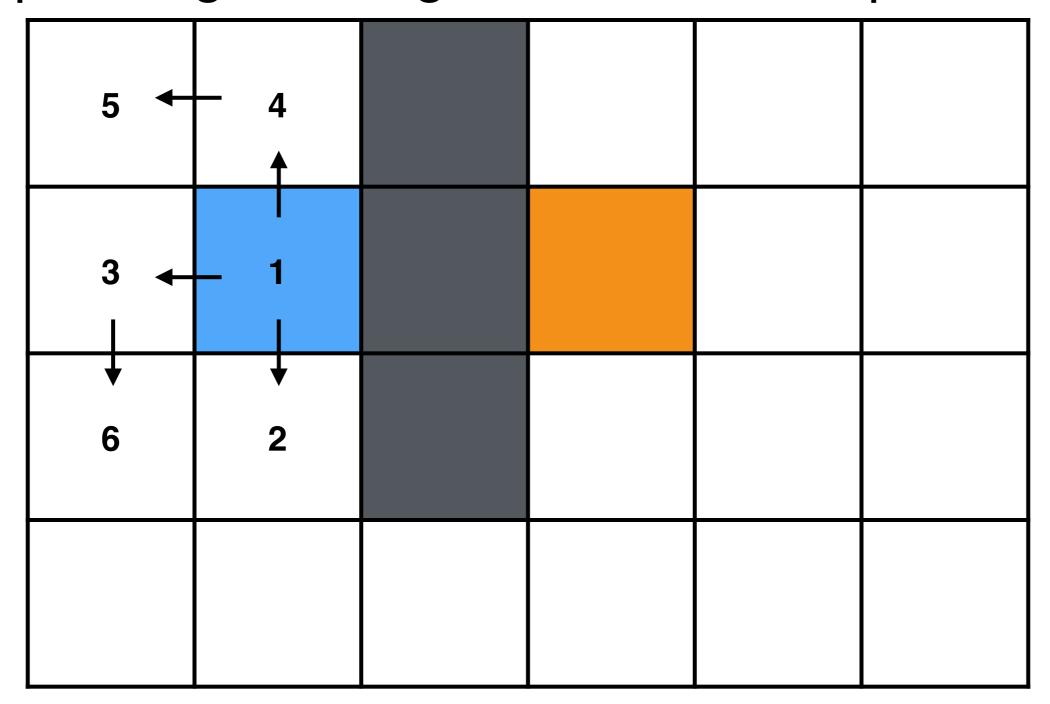
• Q=[2,3,4,5]

• Expand neighbours right, down, left, and, up



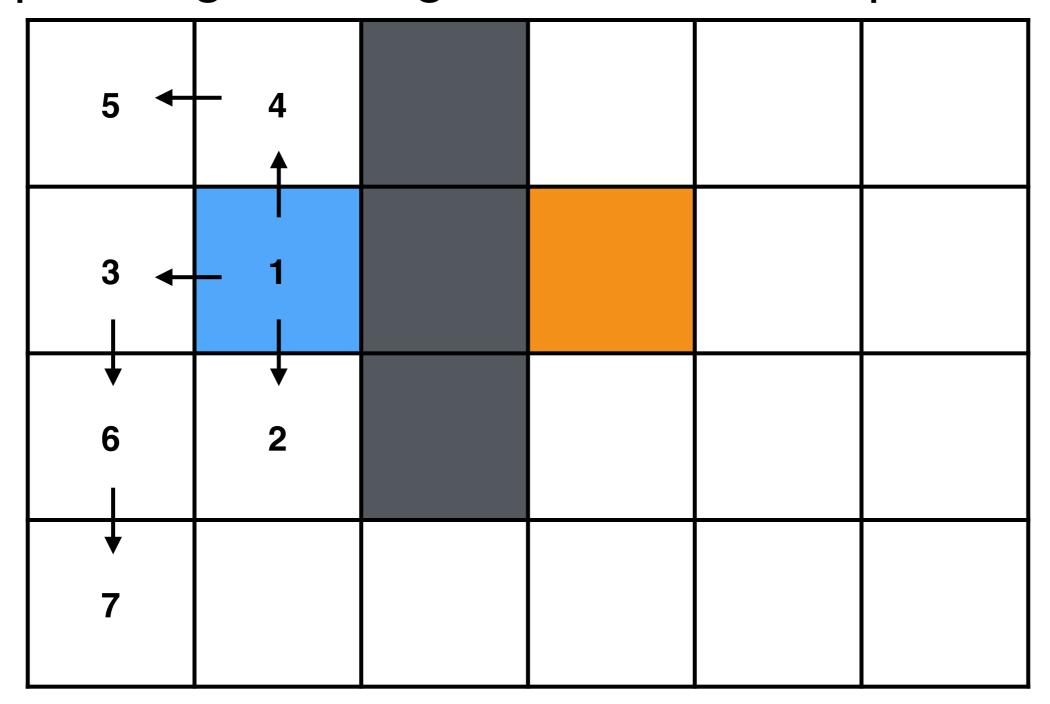
• Q=[2,3,5]

• Expand neighbours right, down, left, and, up



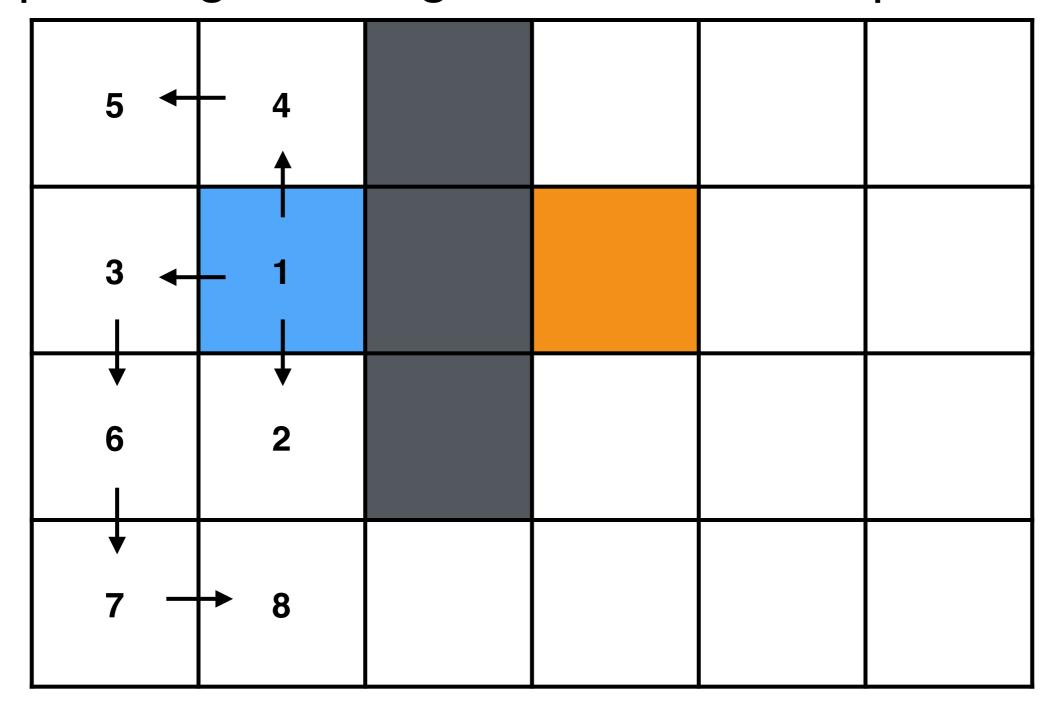
• Q=[2,3,6]

• Expand neighbours right, down, left, and, up



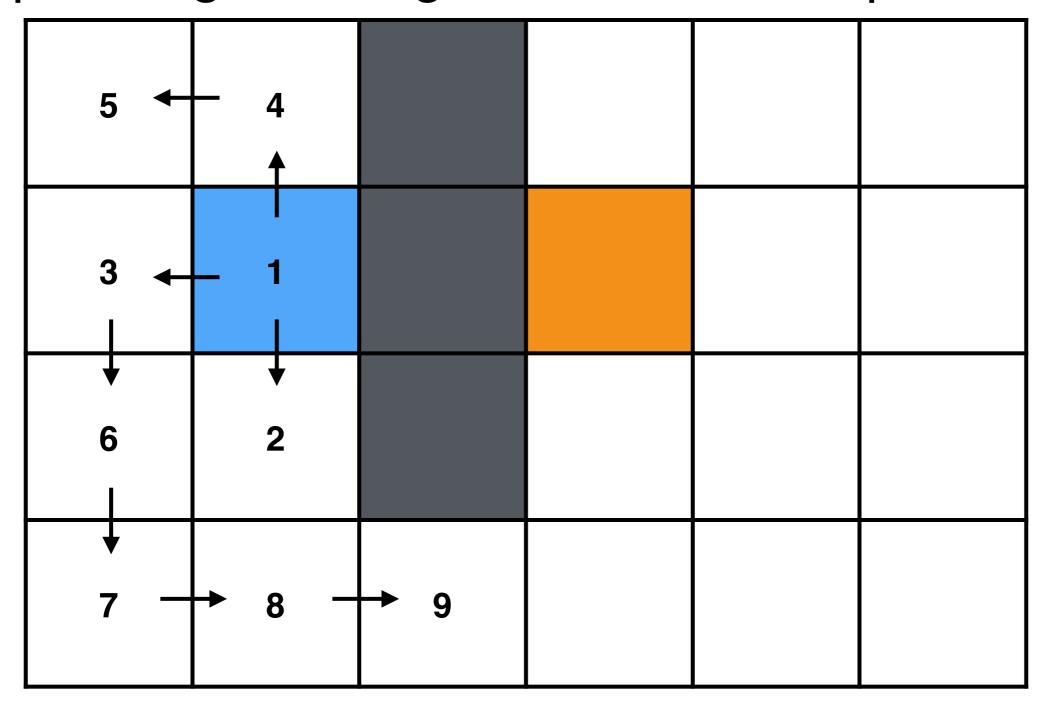
• Q=[2,6,7]

• Expand neighbours right, down, left, and, up



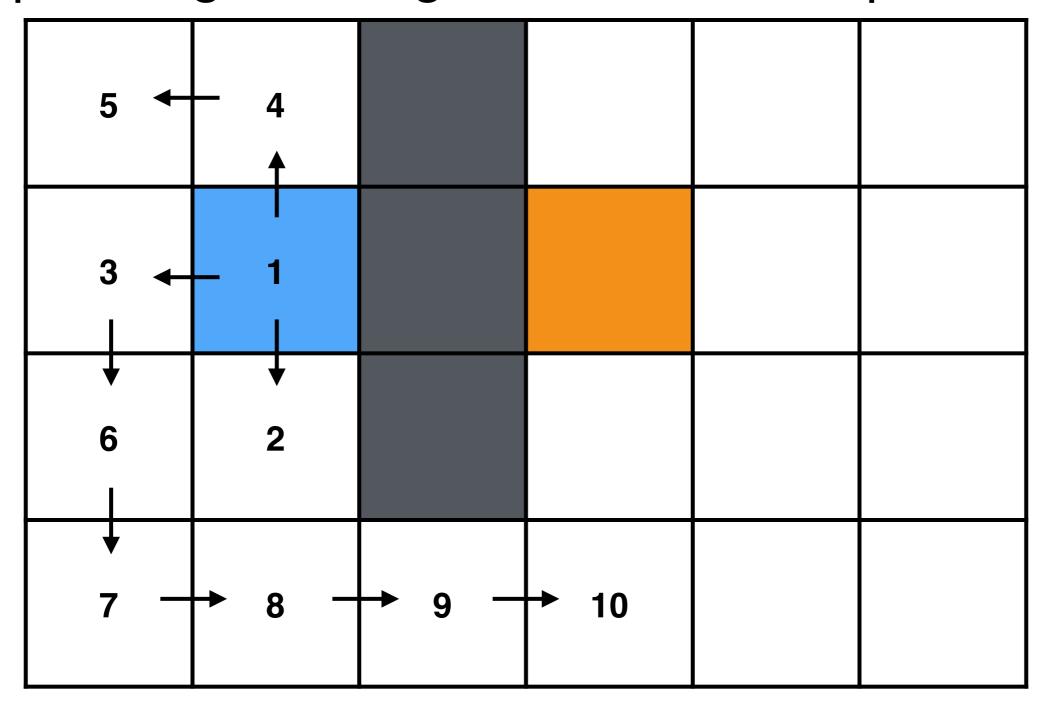
• Q=[2,7,8]

• Expand neighbours right, down, left, and, up



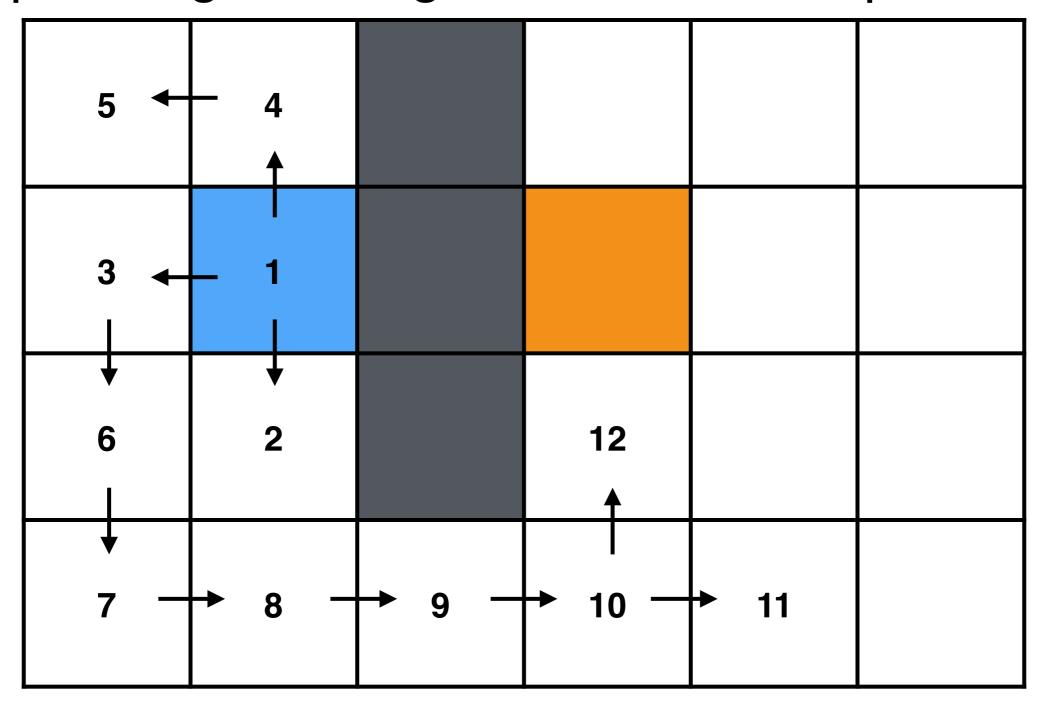
• Q=[2,8,9]

• Expand neighbours right, down, left, and, up



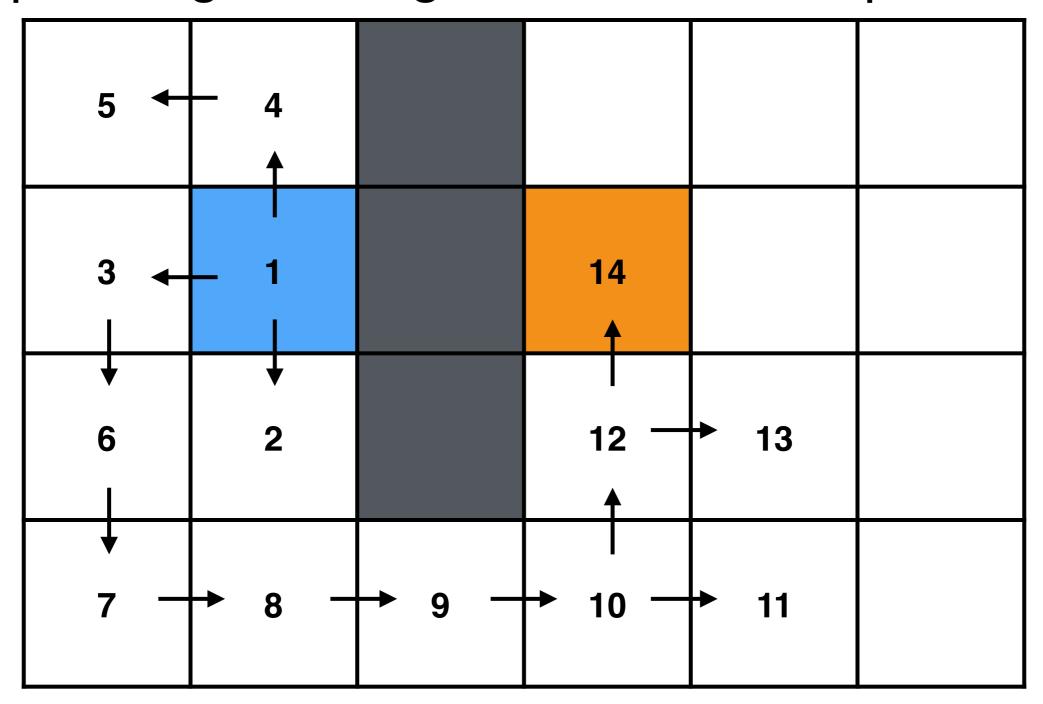
• Q=[2,9,10]

• Expand neighbours right, down, left, and, up



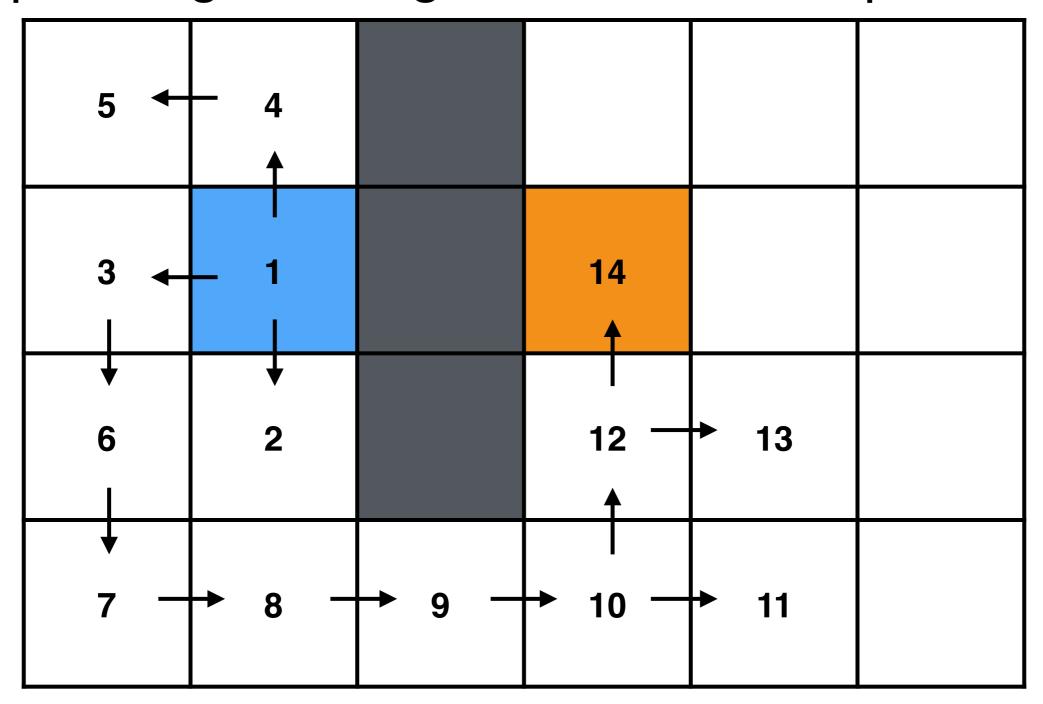
• Q=[2,10,11,12]

Expand neighbours right, down, left, and, up



• Q=[2,11,12,13,14]

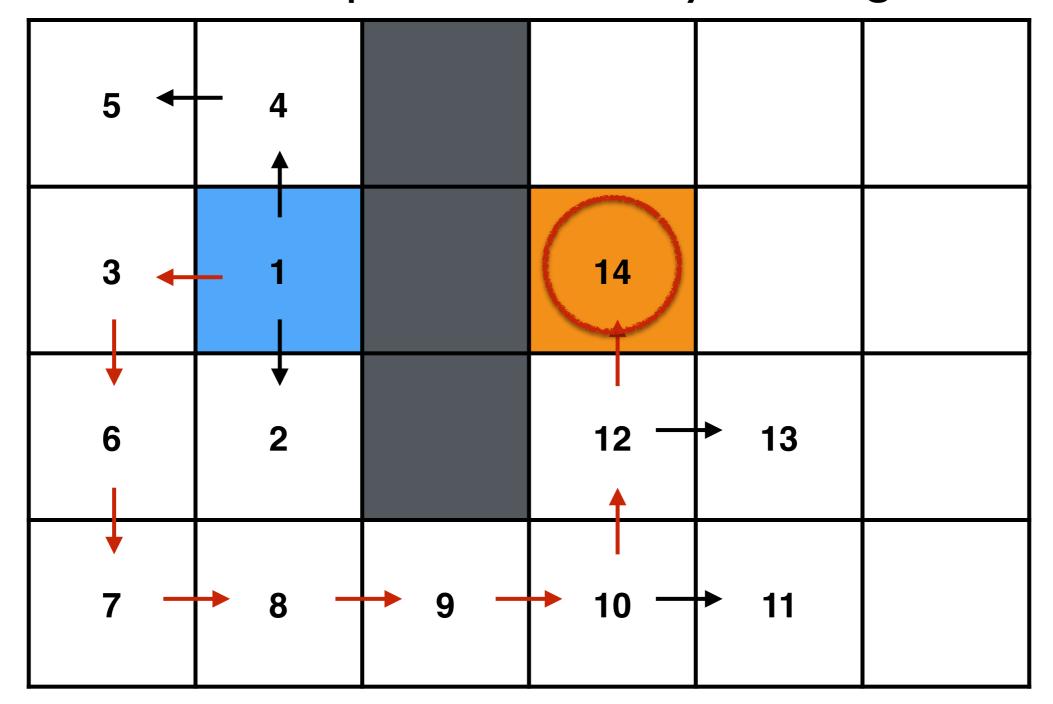
Expand neighbours right, down, left, and, up



• Q=[2,11,13,14] GOAL!!!

#### Depth-First Search

Not the shortest path, can be very winding



So... no advantage over breadth-first... whats the point?

#### Tree Search vs. Graph Search

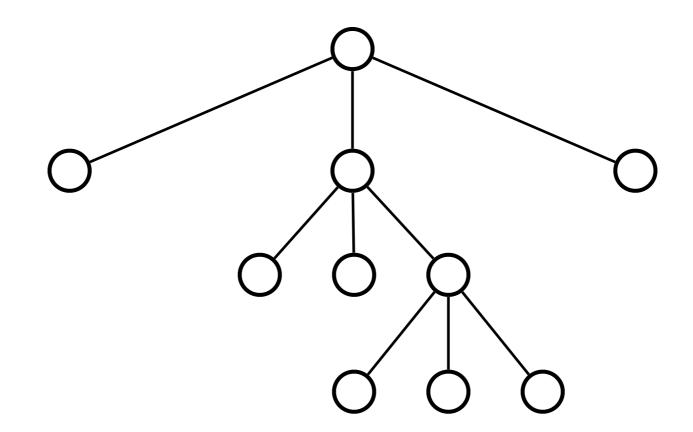
- Tree Search Expand to <u>all</u> neighbours
- Graph Search Expand to neighbours not yet visited

We employed graph search in the examples

- Tree search can have same state multiple times in tree
- Graph search avoids redundant paths and infinite loops
  - Infinite loops make search incomplete even for finite states
  - Tree can avoid loops by comparing to states along path

#### Depth-First Tree Search

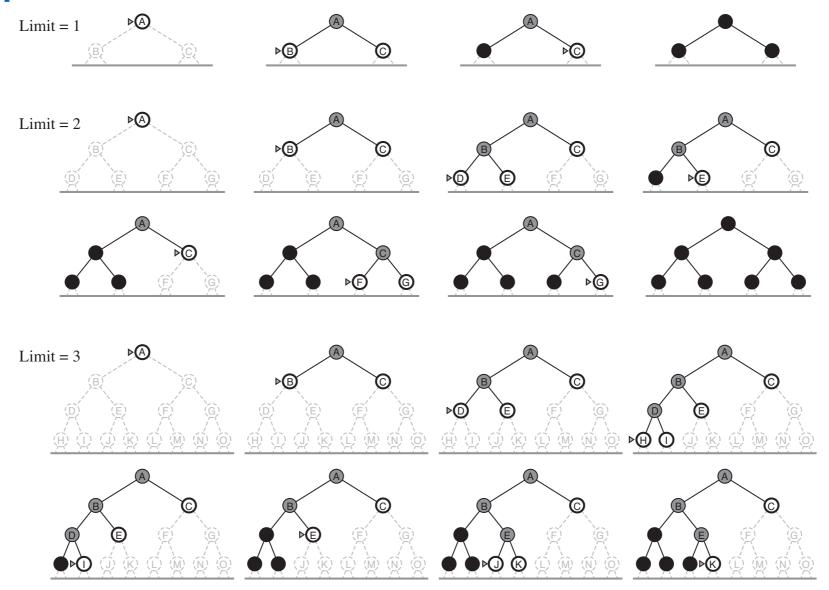
Depth-first tree search is memory efficient



- Only need to store current path and sibling nodes
- Avoid having to store huge tree

## Iterative Deepening

Use depth-first tree with incremental max depth



- Useful for problems with large branching factor
- Low memory usage and finds shortest path

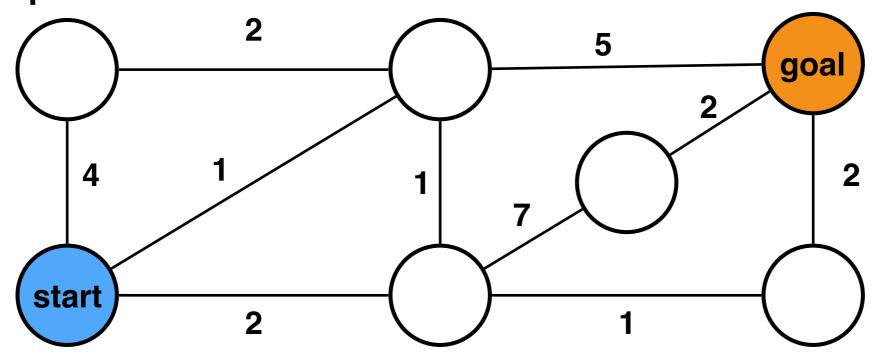
## Optimal Planning

- Not all actions or paths may have equal cost
- Loss function

Want to find a path that minimises loss

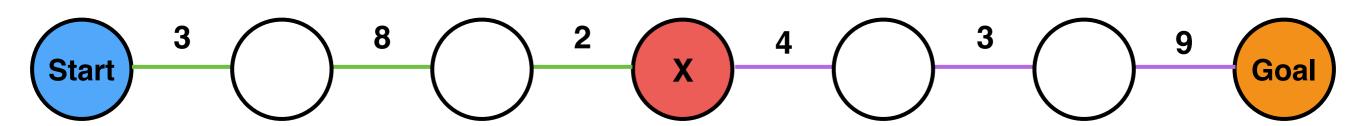
$$\arg\min_{\tau} \sum_{\tau} l(x, u)$$

Example:



#### Cost-to-come and Cost-to-go

Divide total cost into two parts:



#### **Cost-to-come**

$$g(x) = 13$$

#### Cost-to-go

$$h^*(x) = 16$$

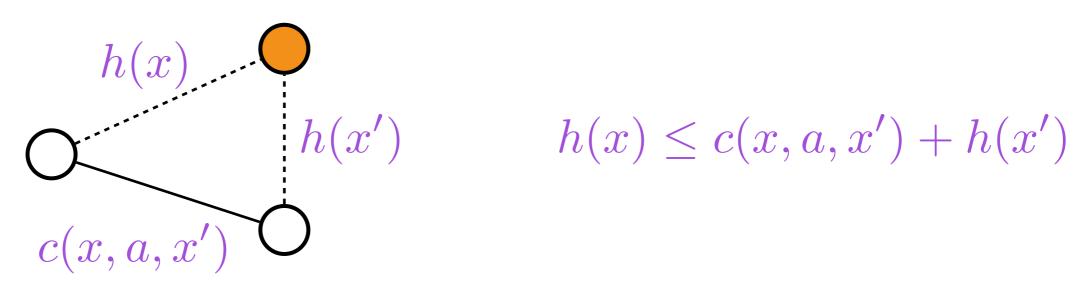
- Cost-to-come is computed as we expand our search
- Cost-to-go is generally not available
- Approximate cost-to-go with a heuristic function

#### What Makes a Good Heuristic?

The heuristic is admissible

$$h(x) \le h^*(x)$$

- Never overestimate the true cost-to-go
- Avoids skipping over the shortest path
- It is consistent/monotonic, follows the triangle inequality



 Needed for finding optimal paths with graph search (or do some extra bookkeeping)

#### What Makes a Good Heuristic?

- Heuristic is often computed by solving an easier problem
  - Additional challenges/constraints only increase distance/cost

- Often use Euclidean distance to goal for shortest path
  - Next node always has a shortcut action for going to the goal
  - Ignore obstacles along the way

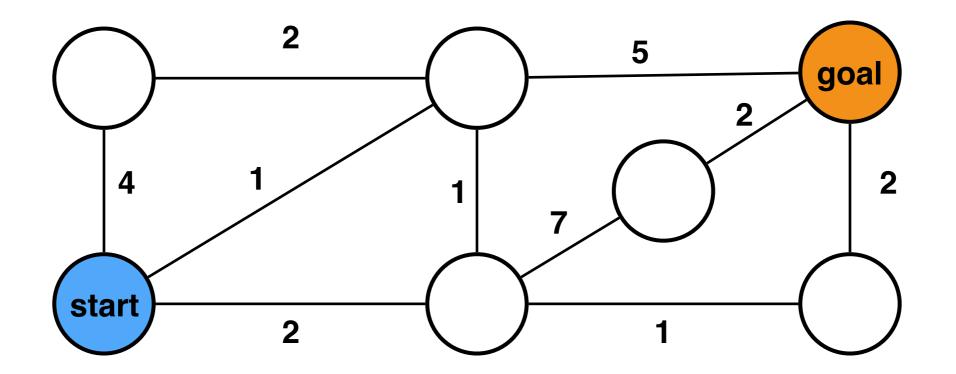
### Search Algorithms

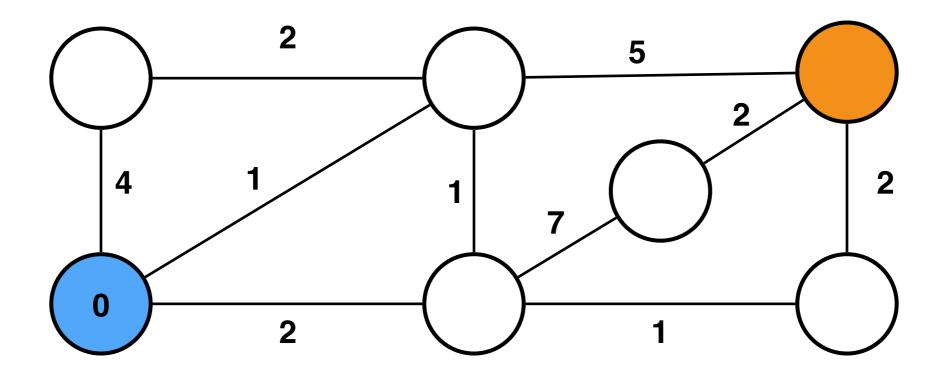
- Organize priority Q according to different values
- Dijkstra's Algorithm

Best-First Search Algorithm

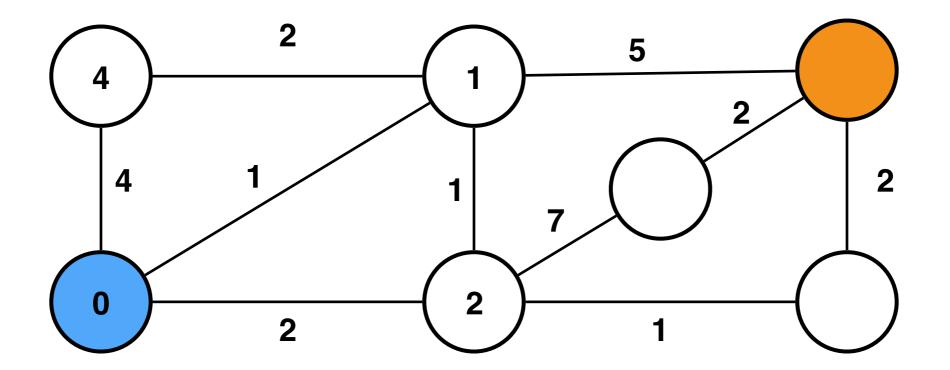
A\* Algorithm ("A star")

$$g(x) + h(x)$$

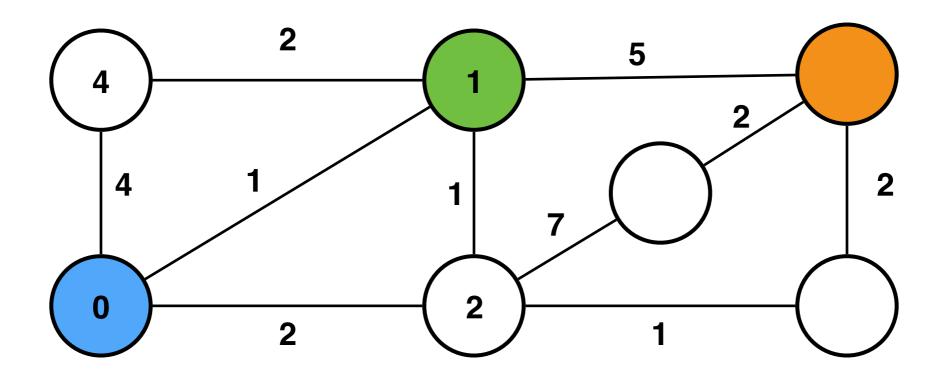




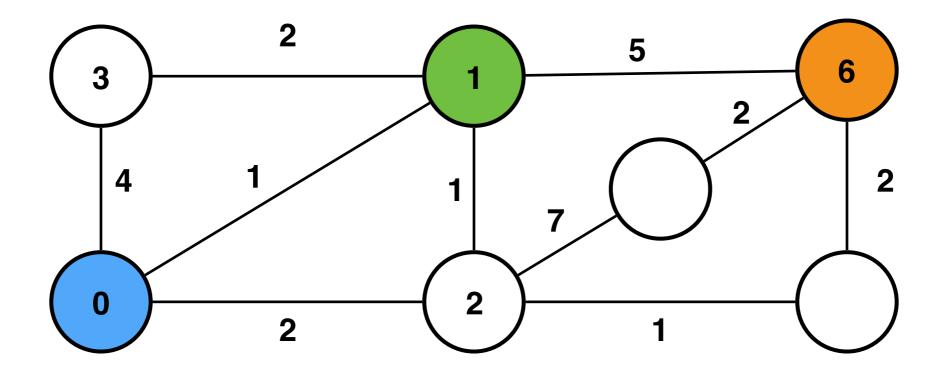
Estimate cost-to-come of neighbours



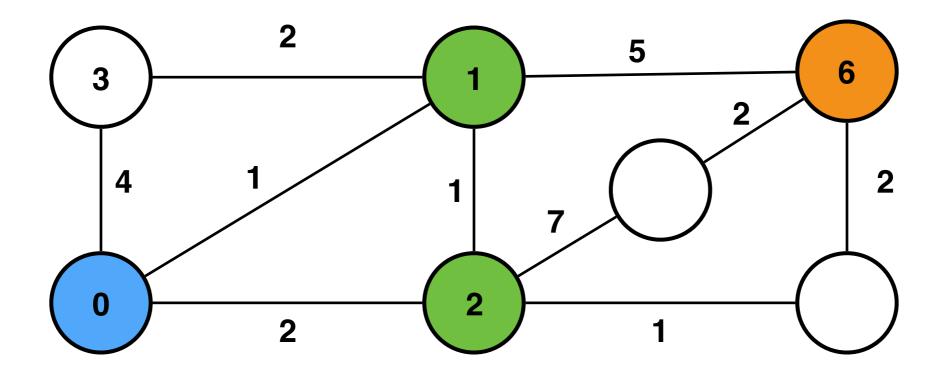
Visit node with lowest cost-to-come (true cost-to-come)



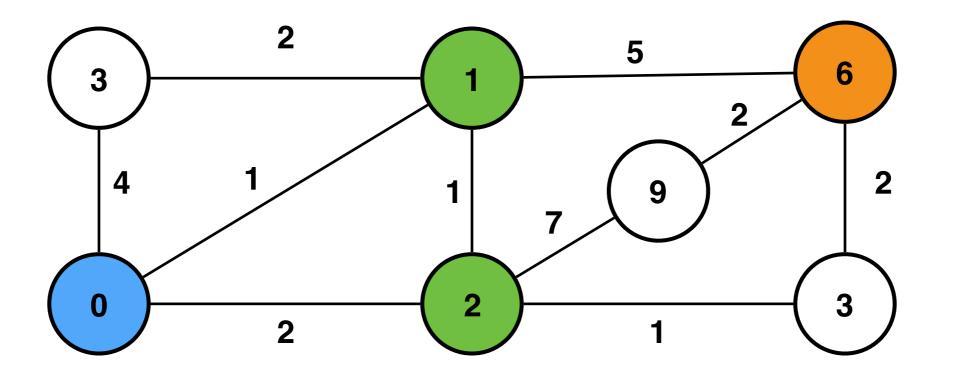
• Estimate cost-to-come of neighbours



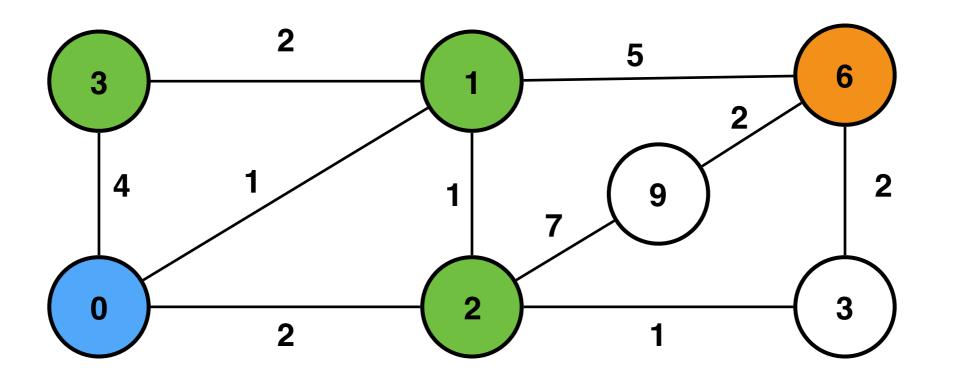
Visit node with lowest cost-to-come



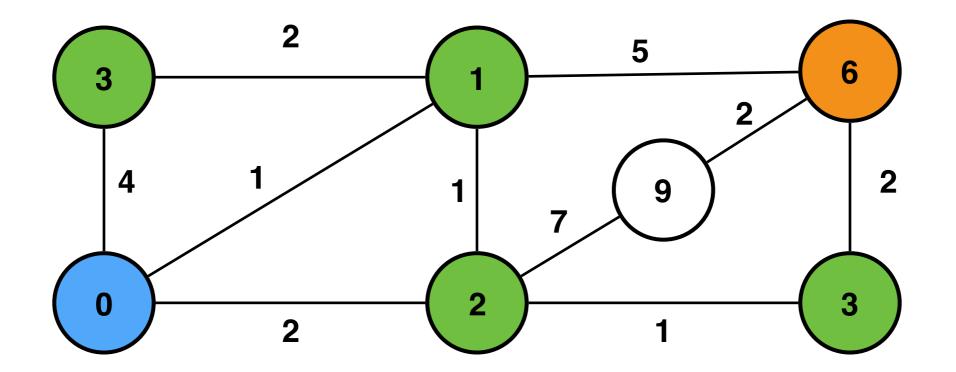
Estimate cost-to-come of neighbours



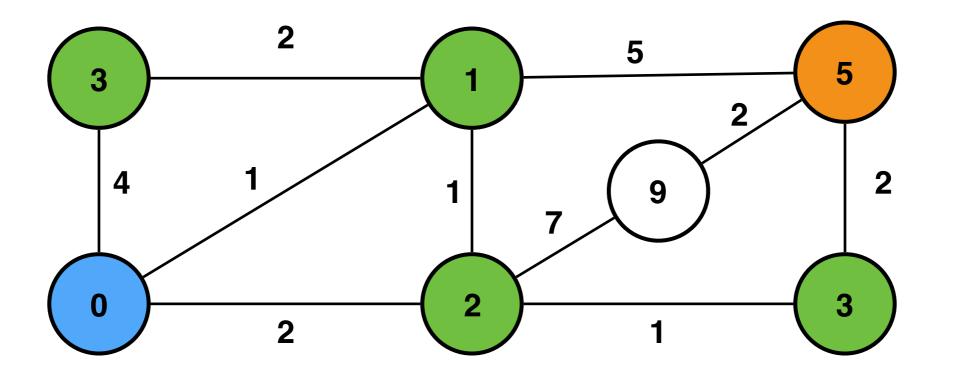
 Visit node with lowest cost-to-come (no unvisited neighbours)



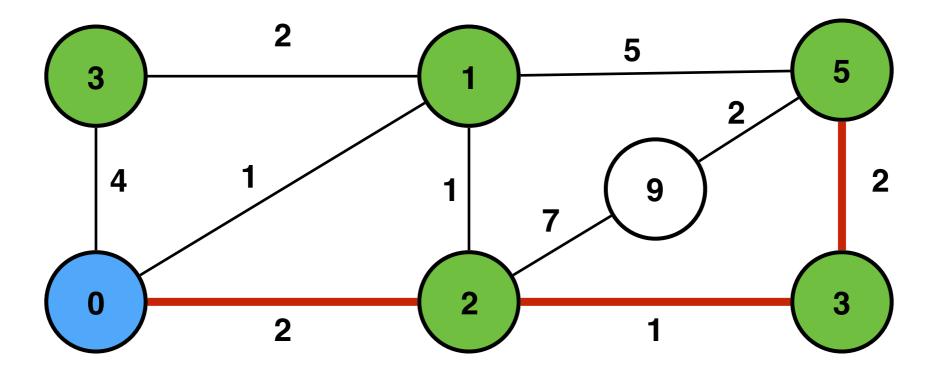
Visit node with lowest cost-to-come

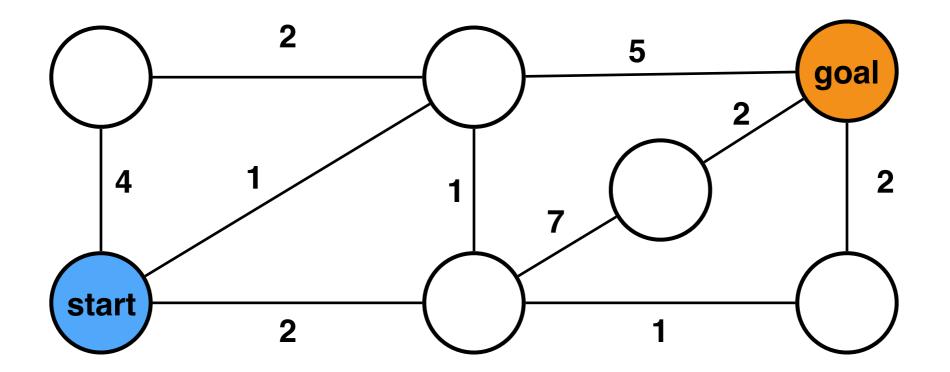


Estimate cost-to-come of neighbours

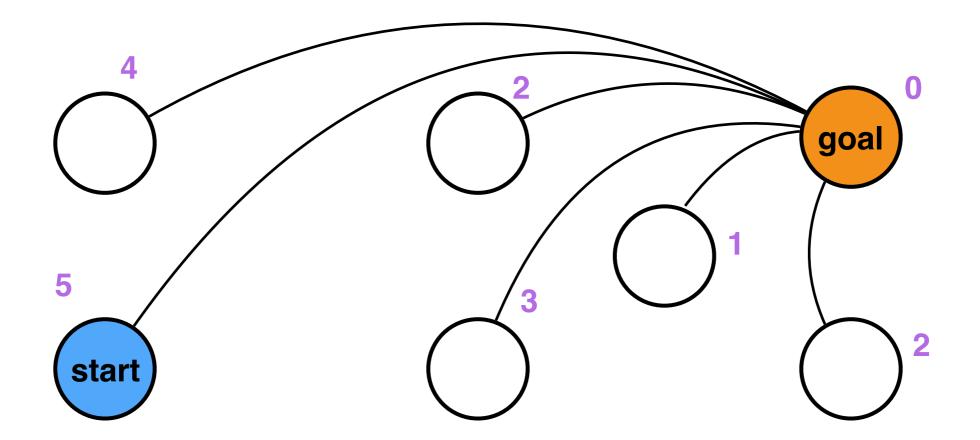


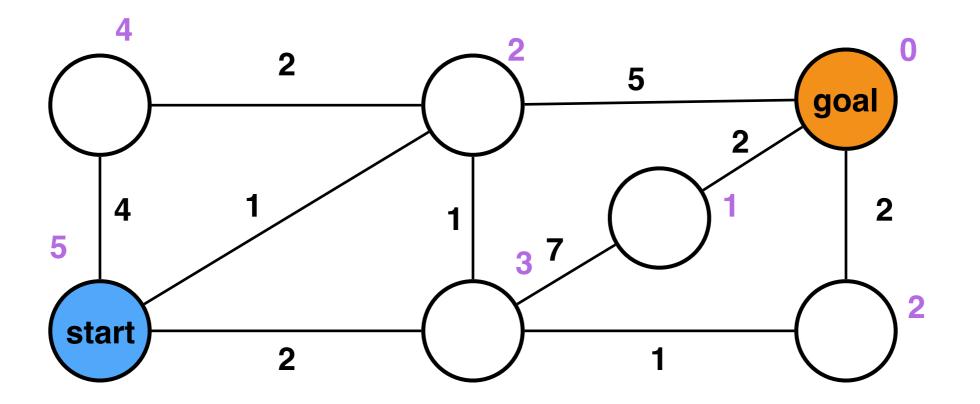
Visit node with lowest cost-to-come GOAL!!!



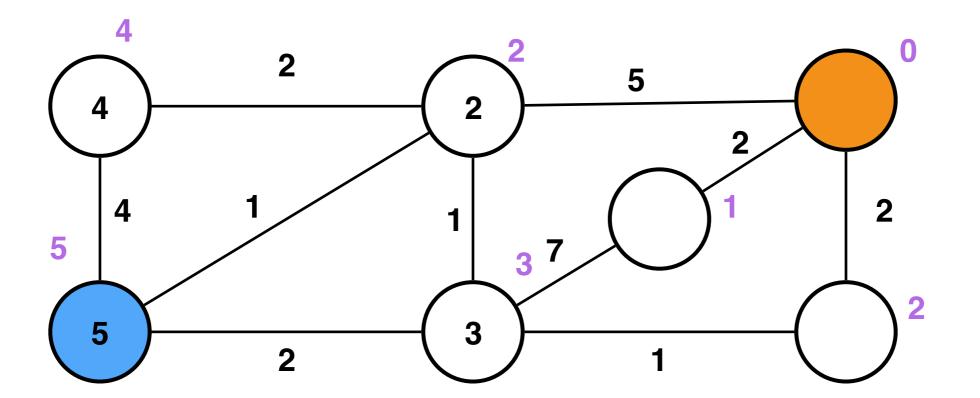


Estimate the cost-to-go (can do this online)

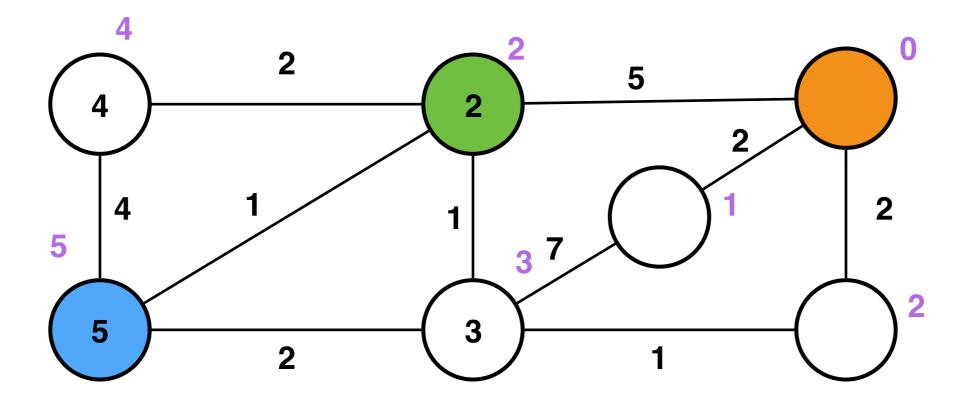




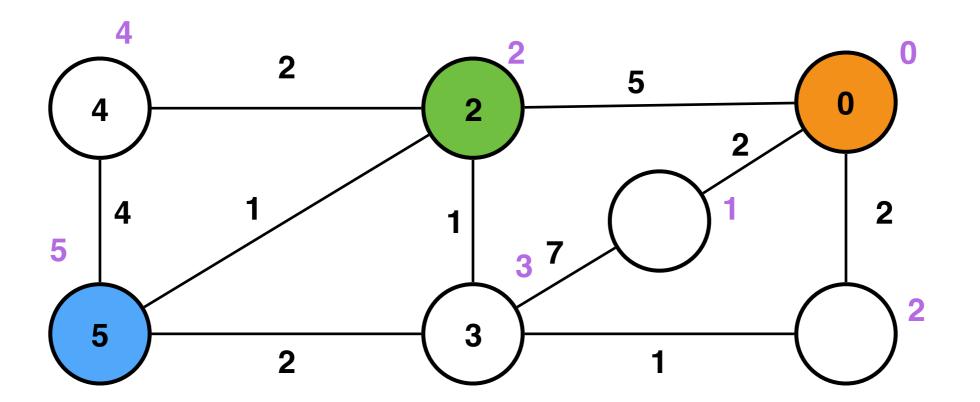
Compute cost-to-go estimates of neighbouring states



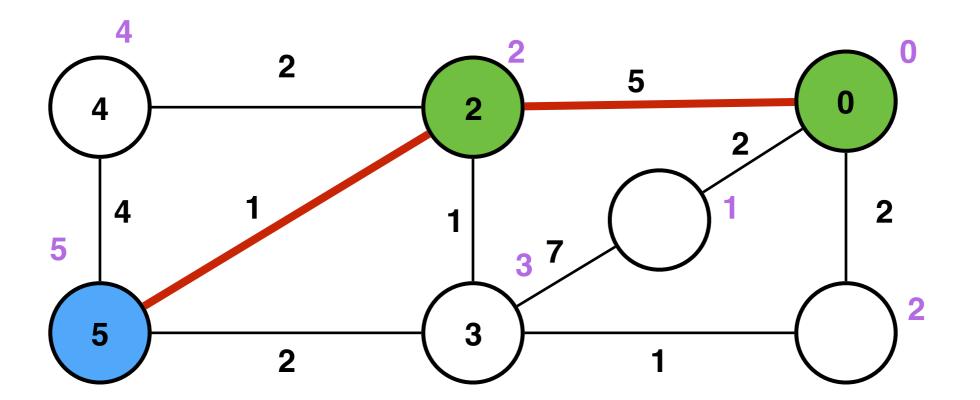
Expand to node closest to goal

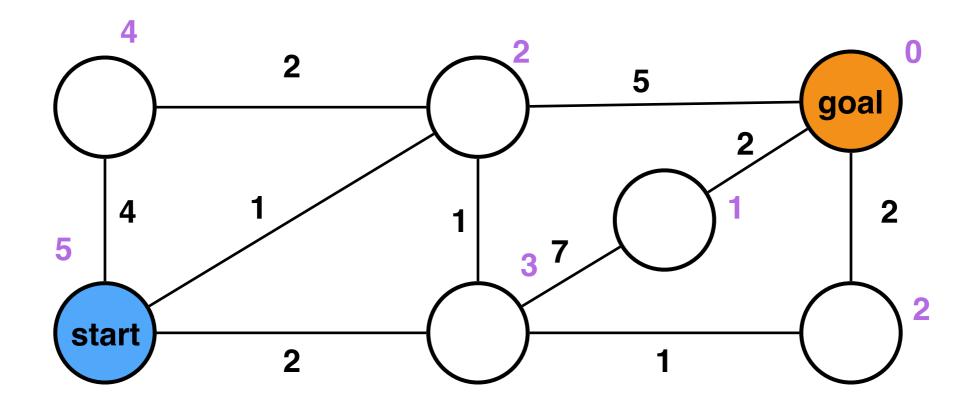


Compute cost-to-go estimates

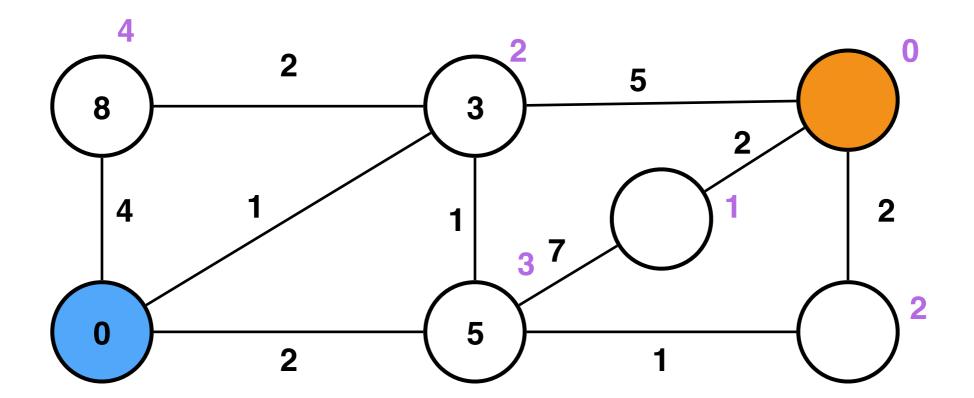


Expand to node closest to goal. GOAL!!! (not cheapest)

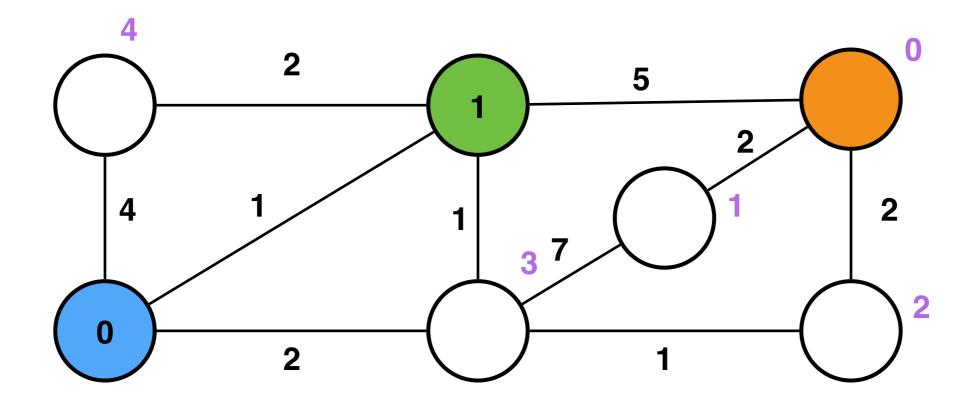




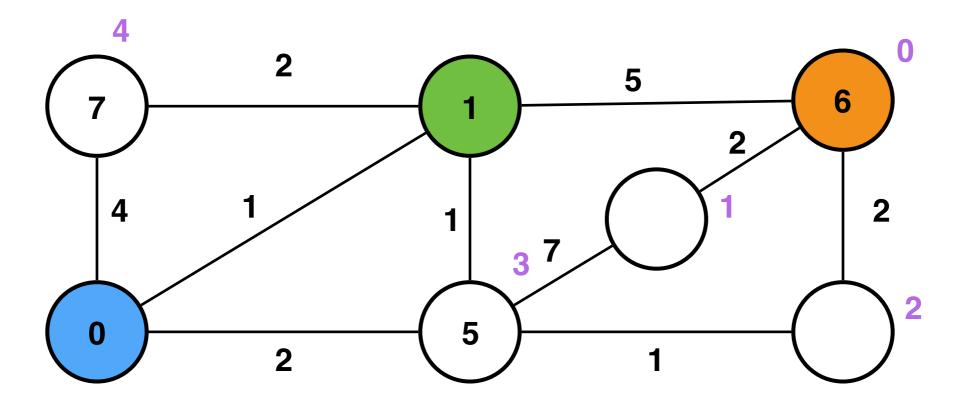
• Estimate cost-to-come + cost-to-go



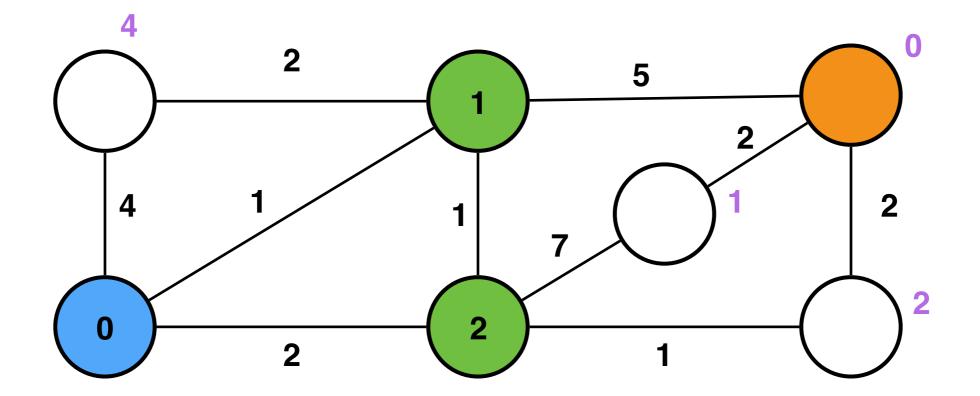
Expand to lowest-value node



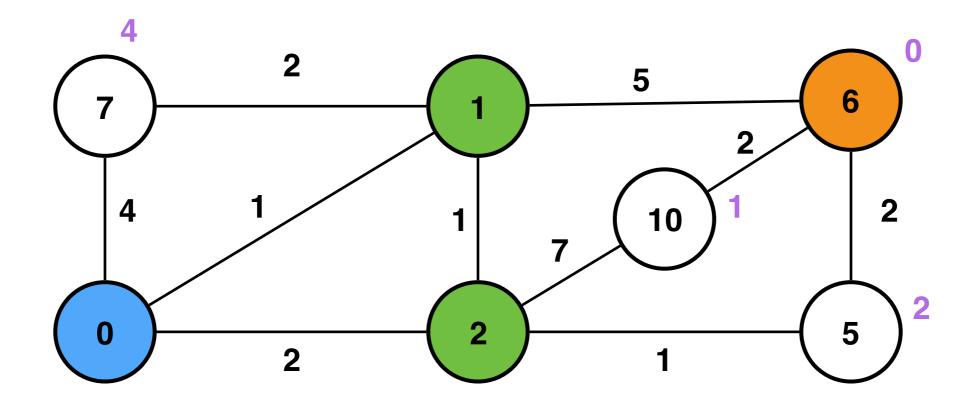
Estimate cost-to-come + cost-to-go



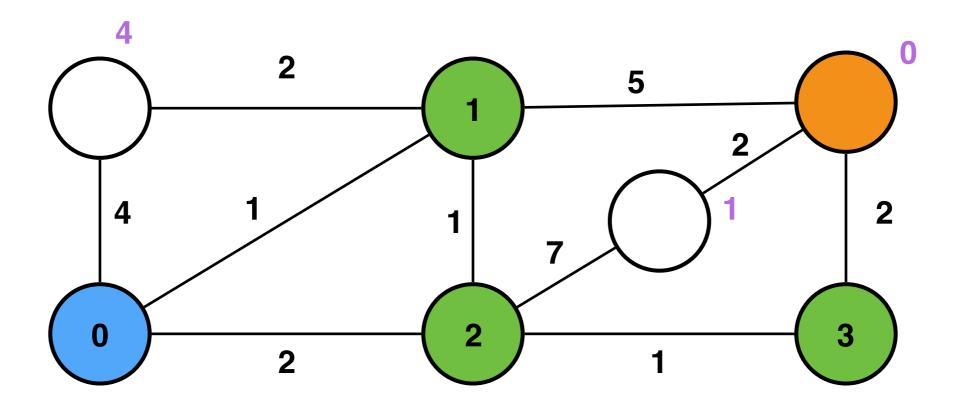
Expand to lowest-value node



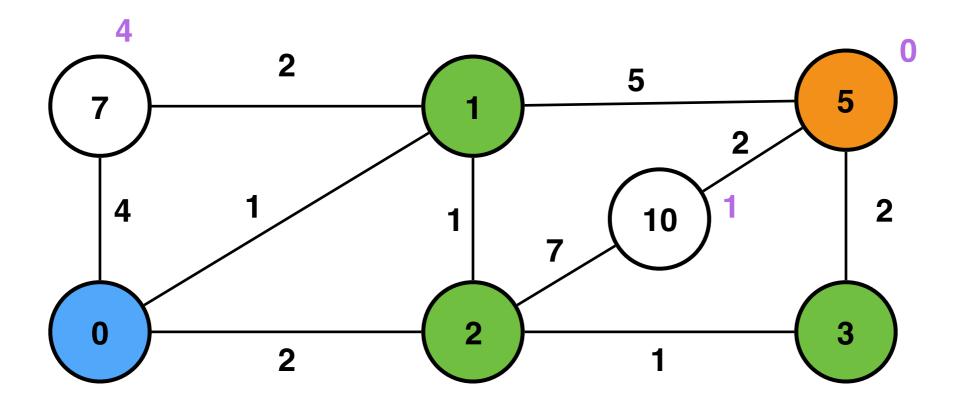
• Estimate cost-to-come + cost-to-go



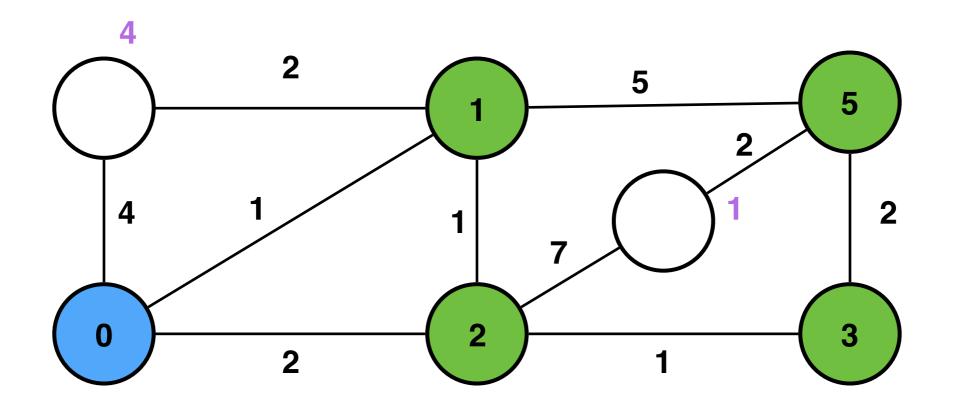
Expand to lowest-value node



Estimate cost-to-come + cost-to-go



Expand to lowest-value node GOAL!!! Shortest path



### Search Algorithms

Dijkstra's Algorithm

Best-First Search Algorithm

A\* Algorithm ("A star")

$$g(x) + h(x)$$

Questions?