

# CS221 Section 3: Search

Intuition and Examples

Vig & Rickard

# Search

How to find shortest path  
to my destination?

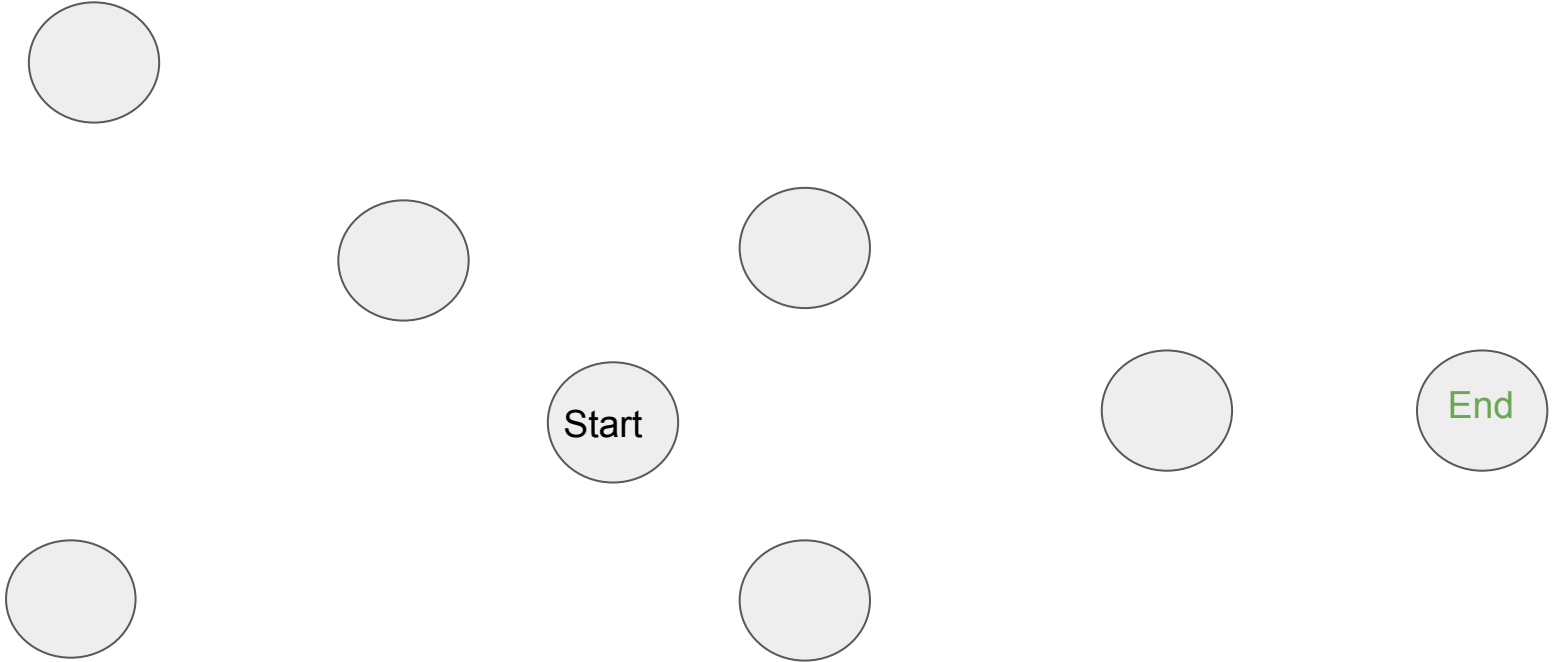


HOME

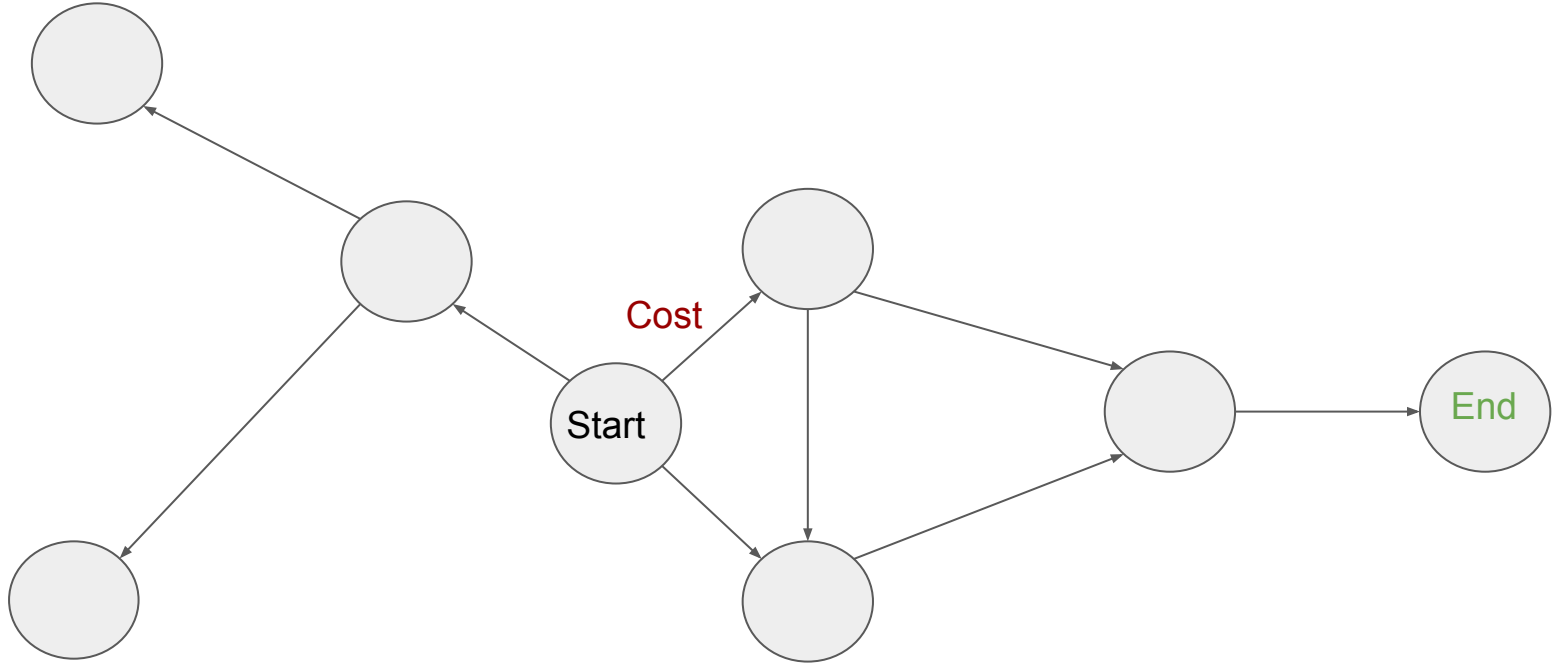


DESTINATION

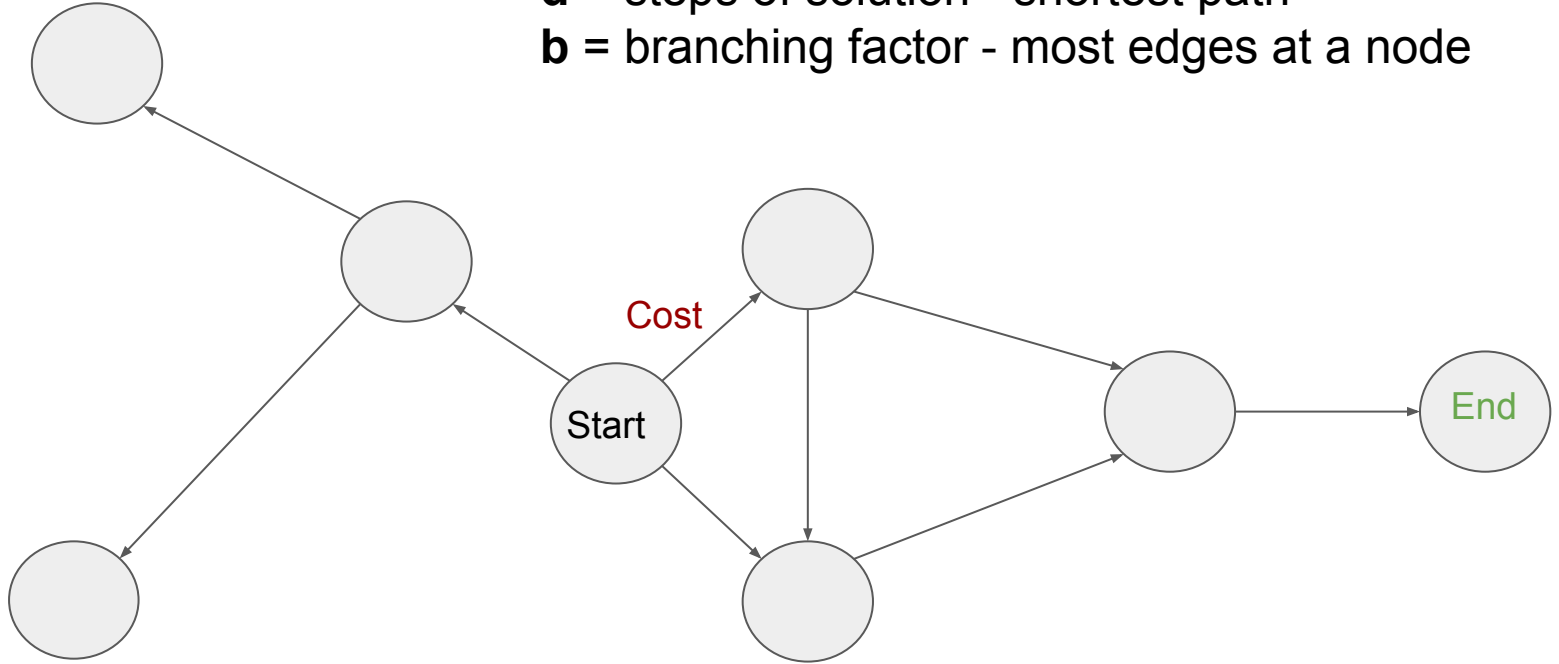
# Search



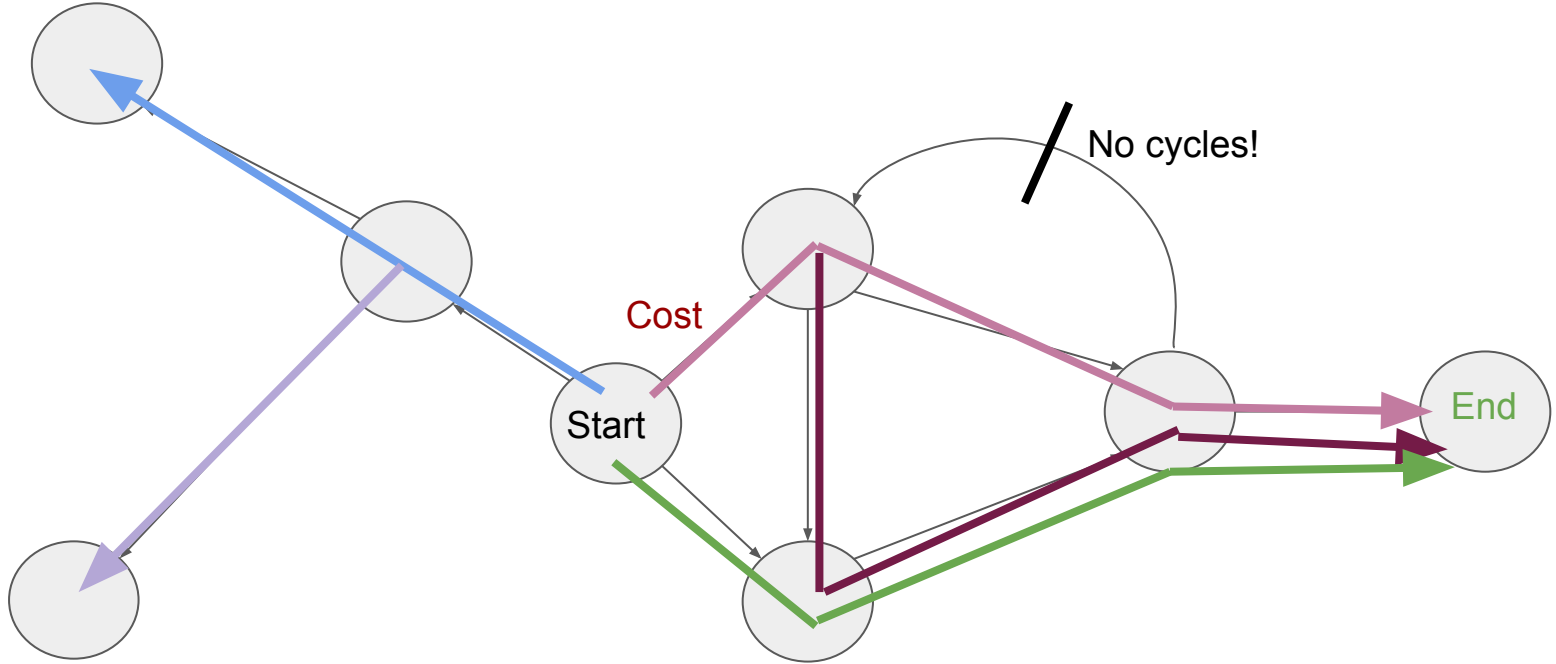
# Search!



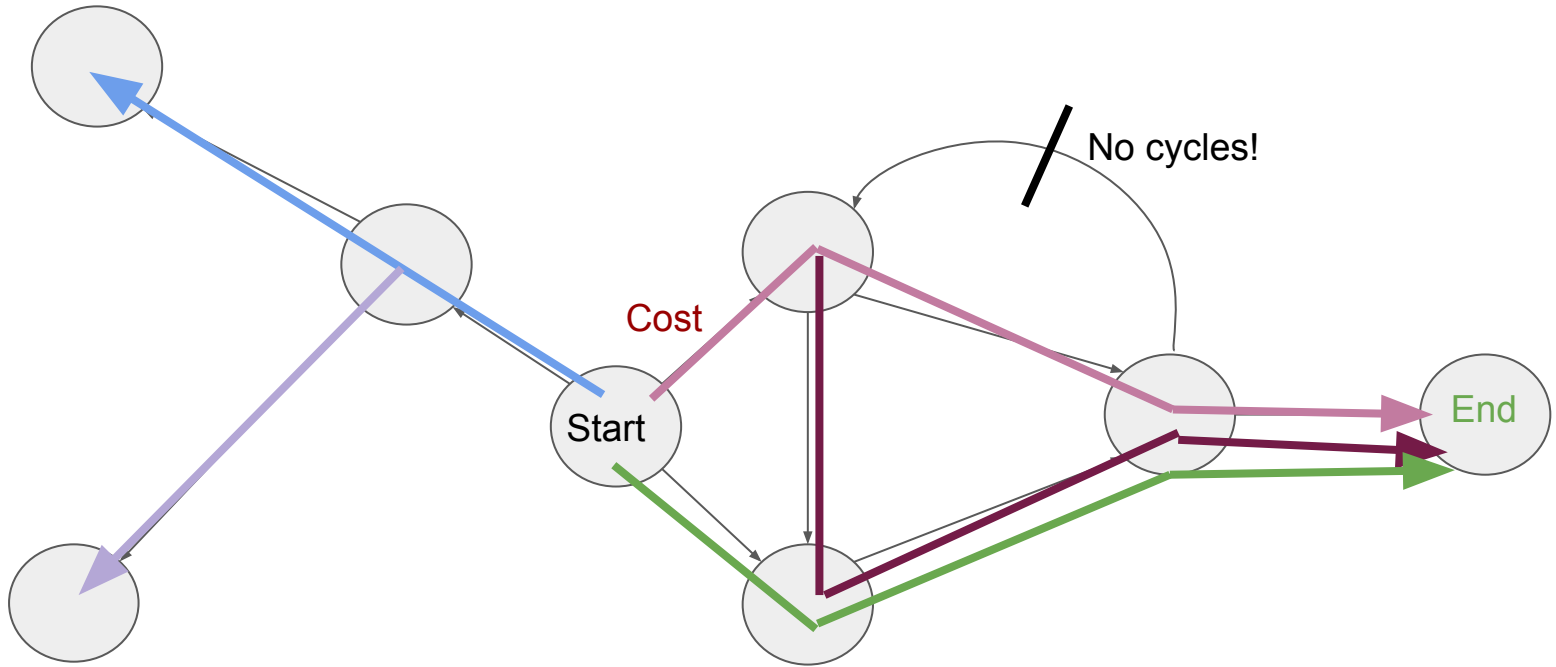
**D** = maximum depth - longest path  
**d** = steps of solution - shortest path  
**b** = branching factor - most edges at a node



Just try them all! Backtracking



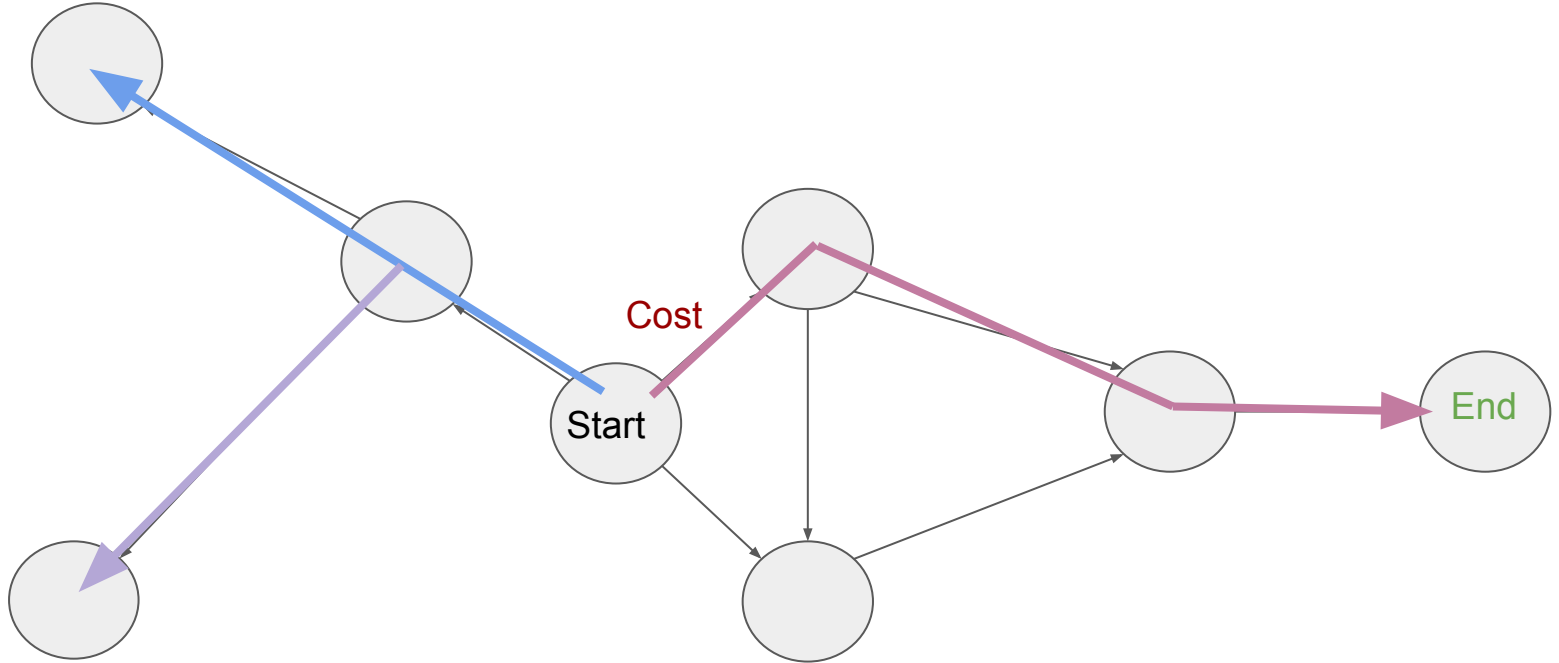
Just try them all! Backtracking



Expensive:  $O(b^D)$

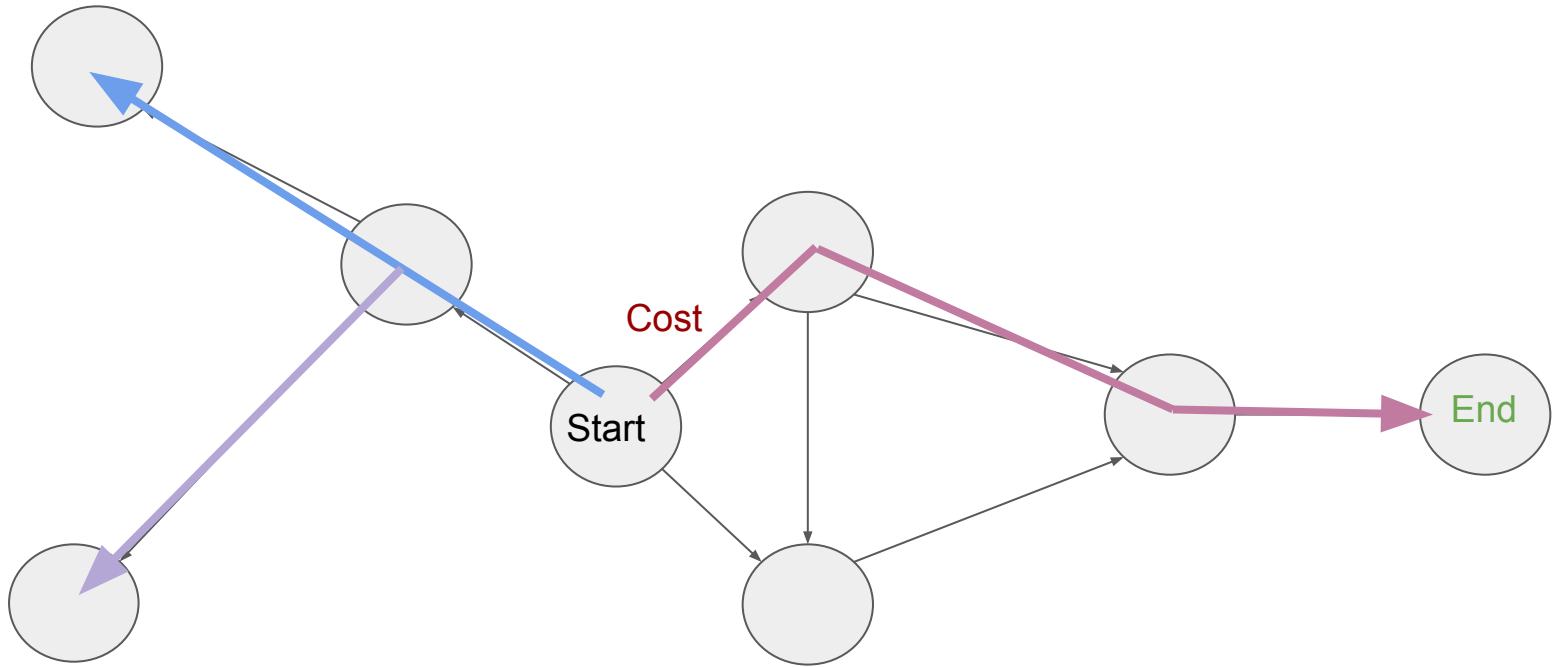
Cheap in space:  $O(D)$

Try them all until you find one! DFS





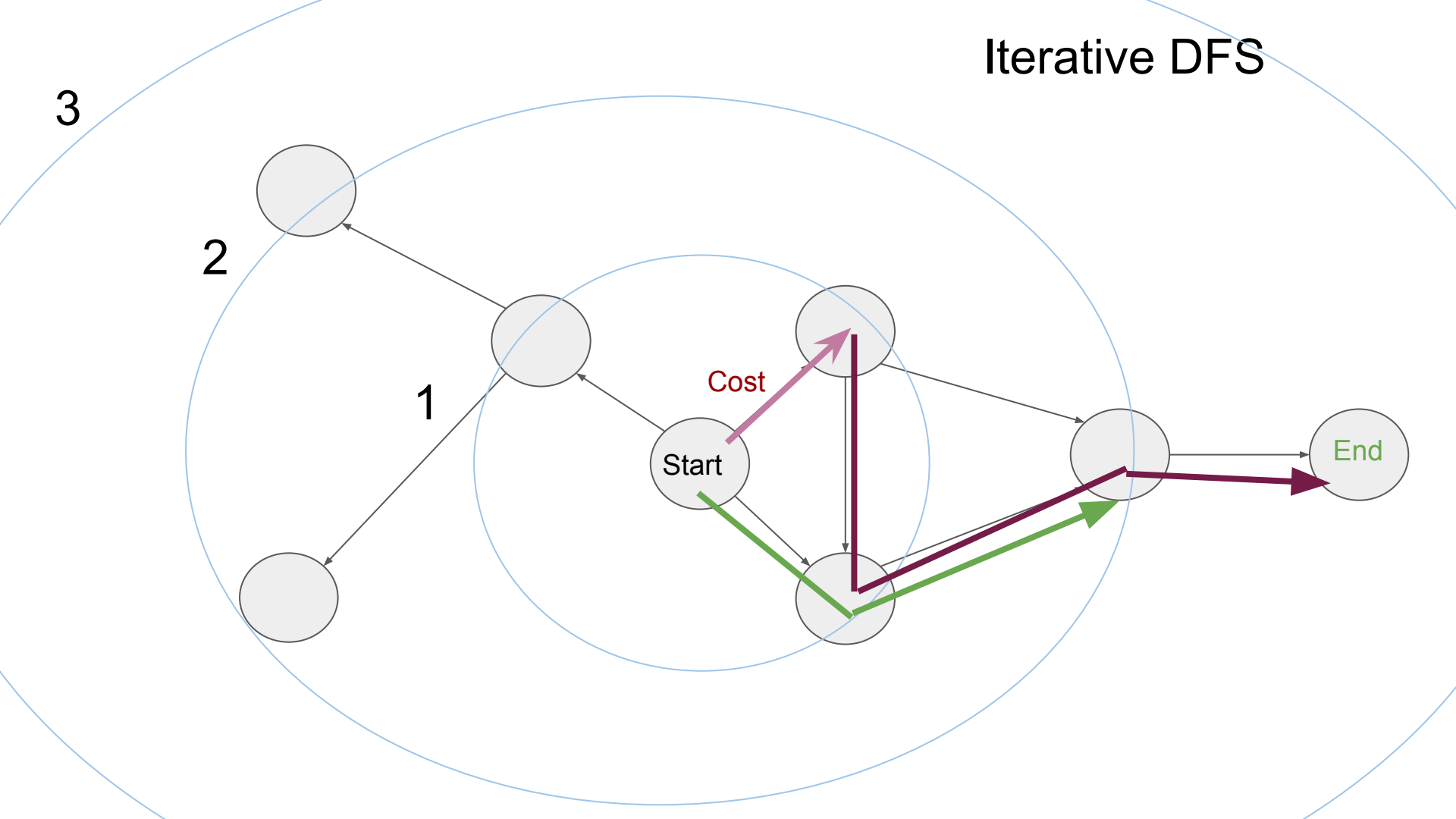
Try them all until you find one! DFS



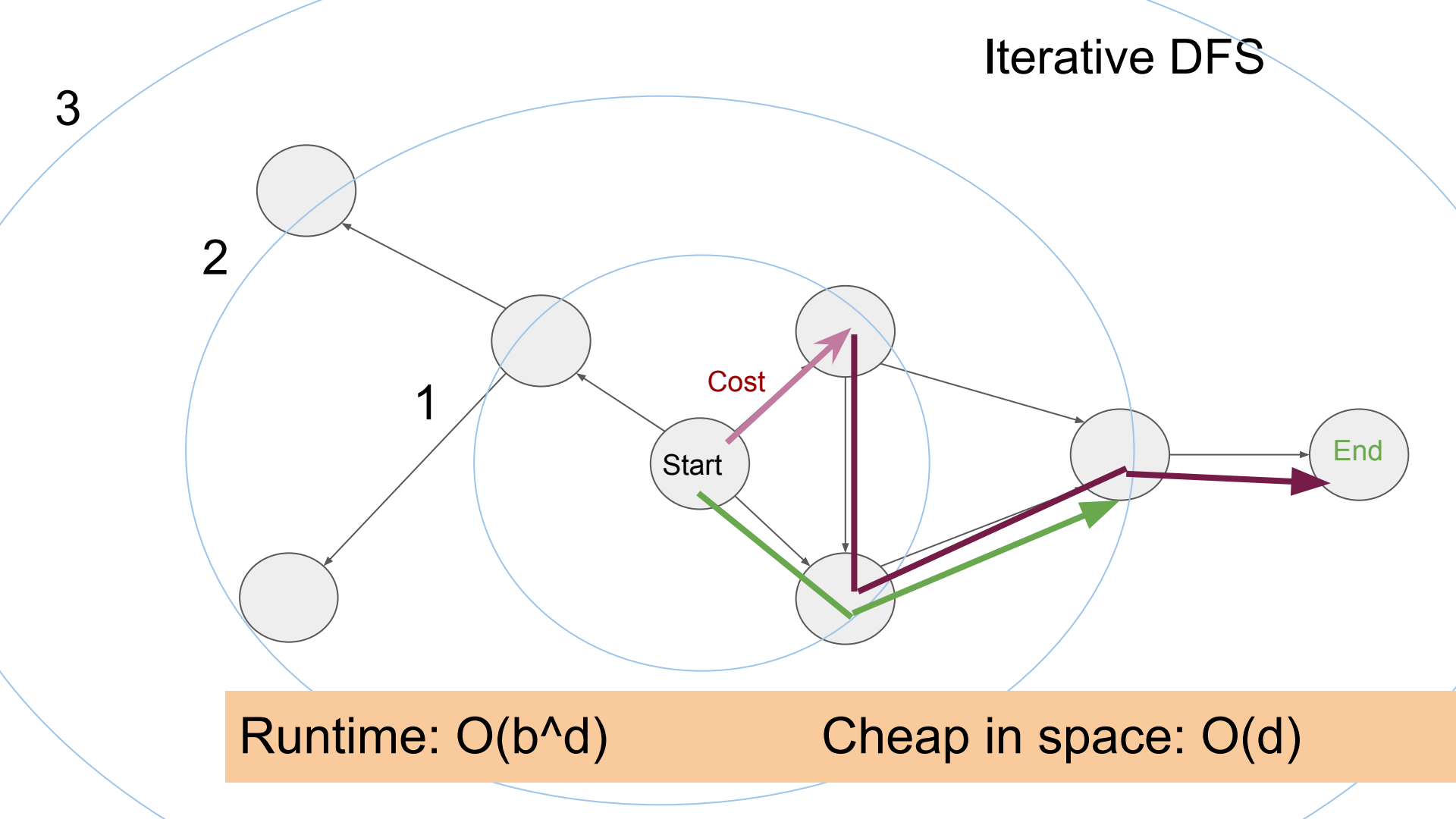
Expensive:  $O(b^D)$

Cheap in space:  $O(D)$

# Iterative DFS



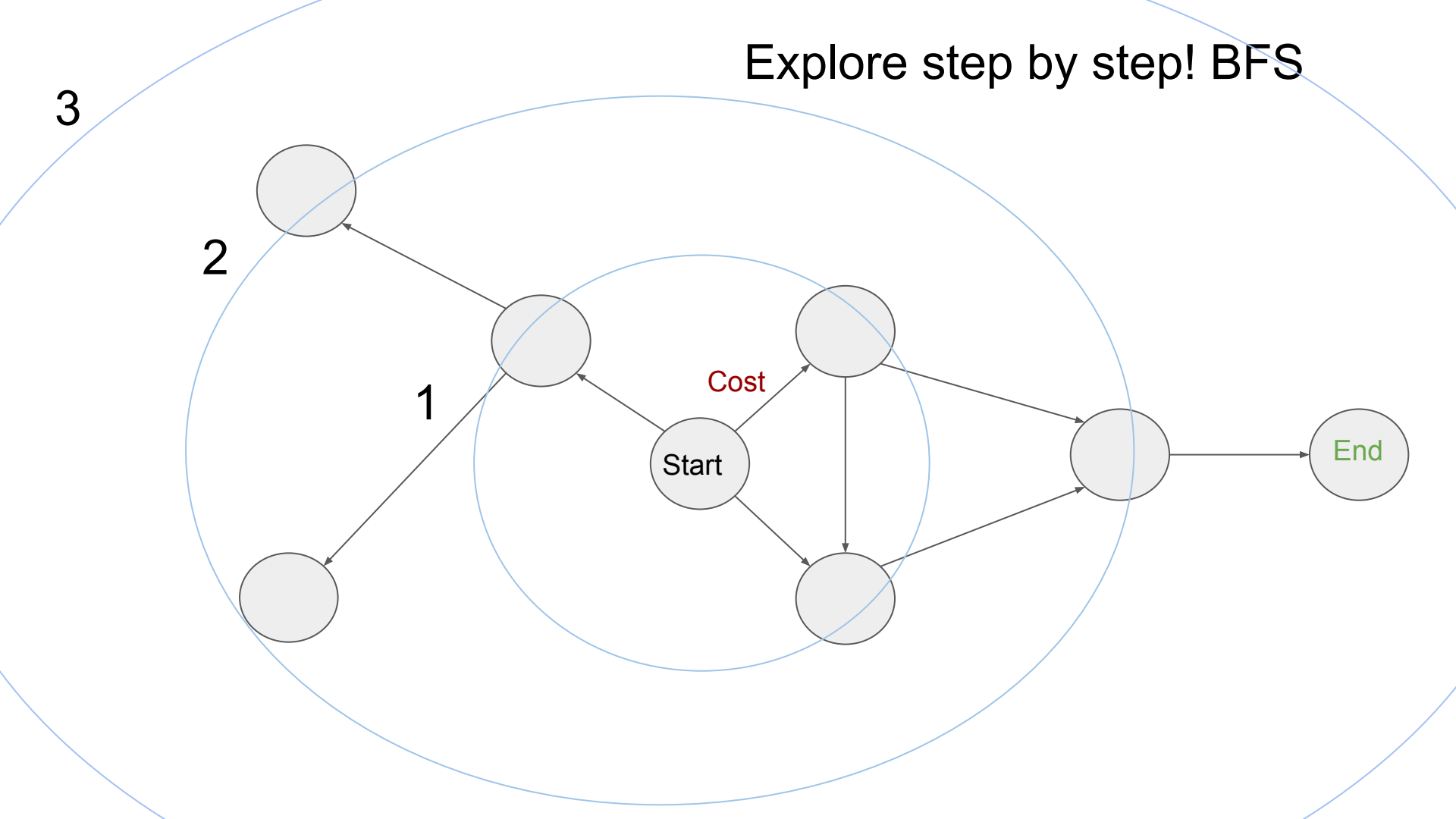
# Iterative DFS



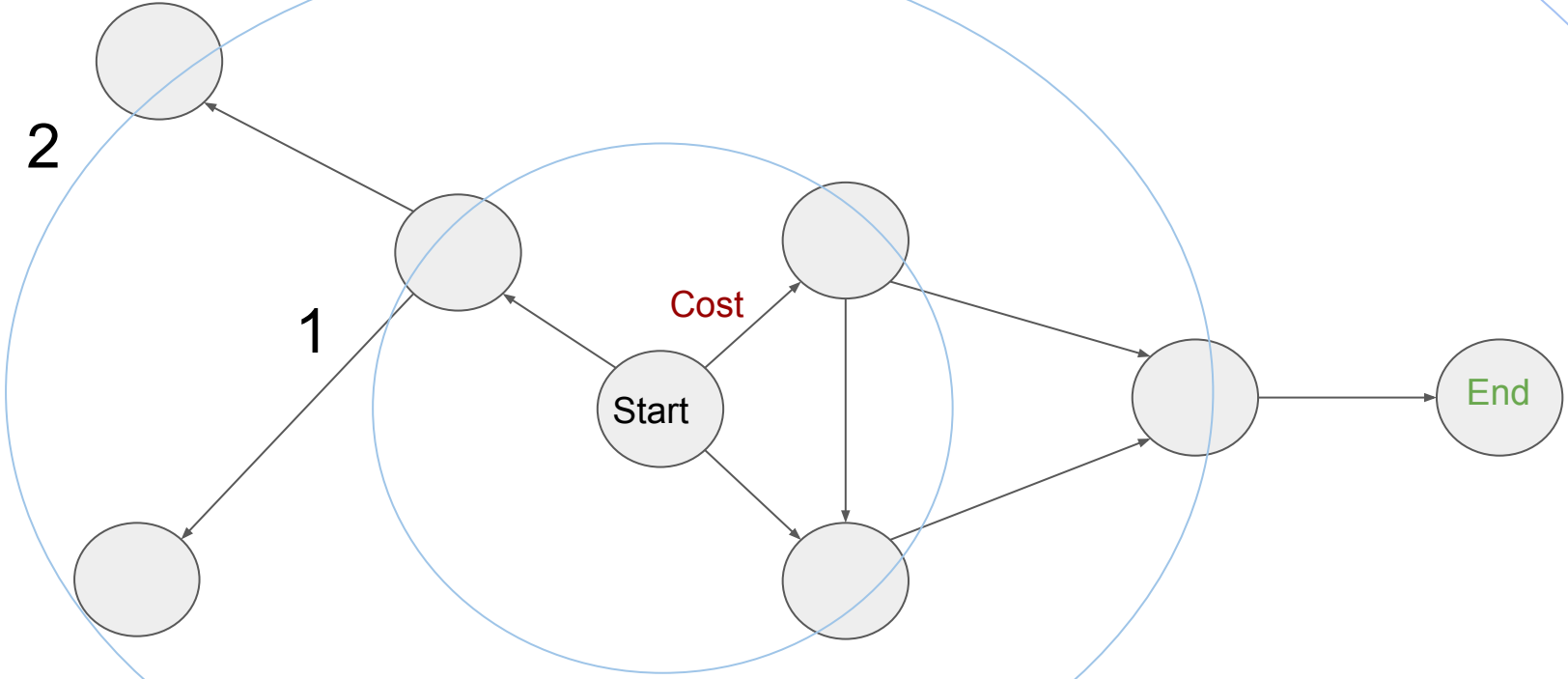
Runtime:  $O(b^d)$

Cheap in space:  $O(d)$

Explore step by step! BFS



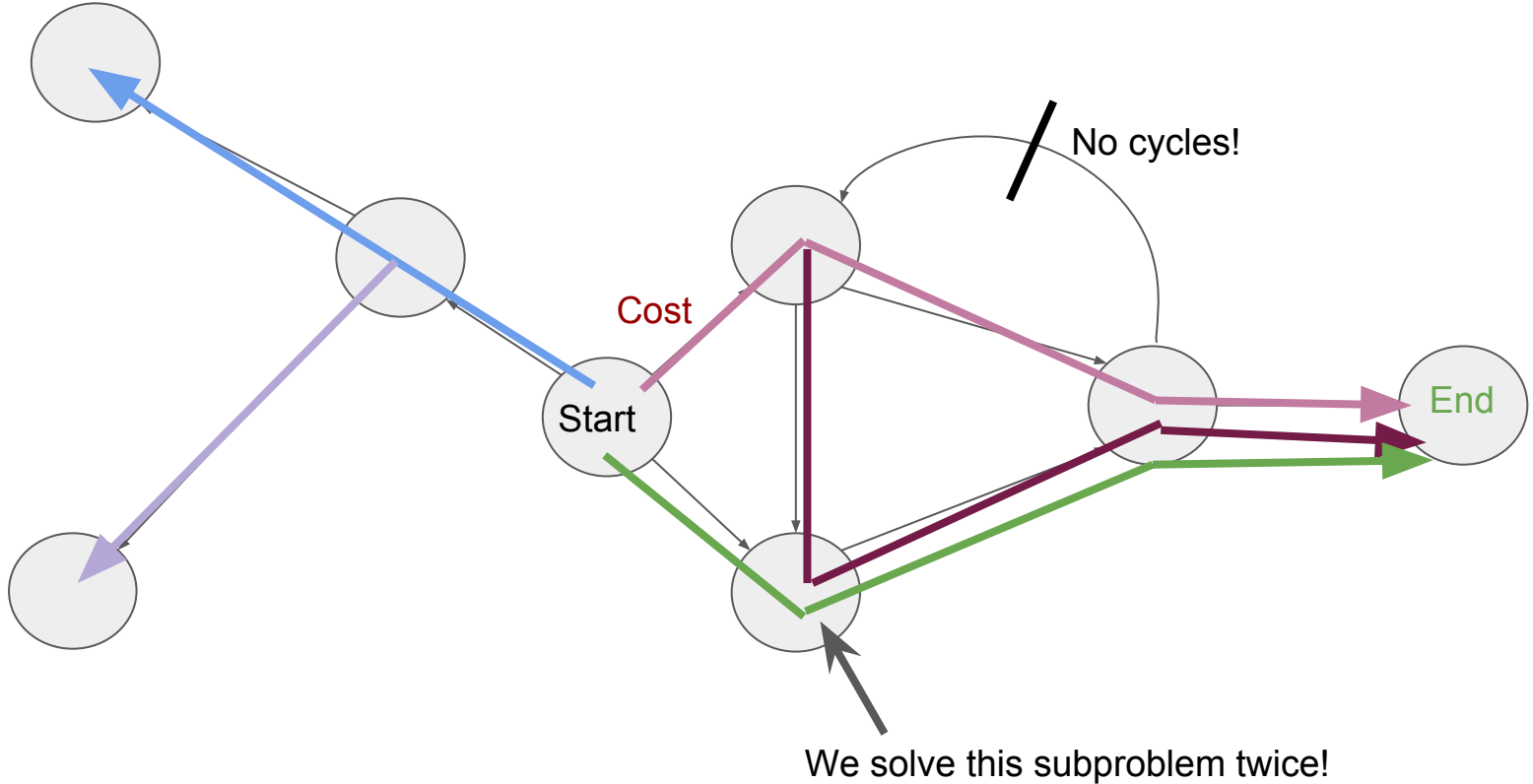
Explore step by step! BFS



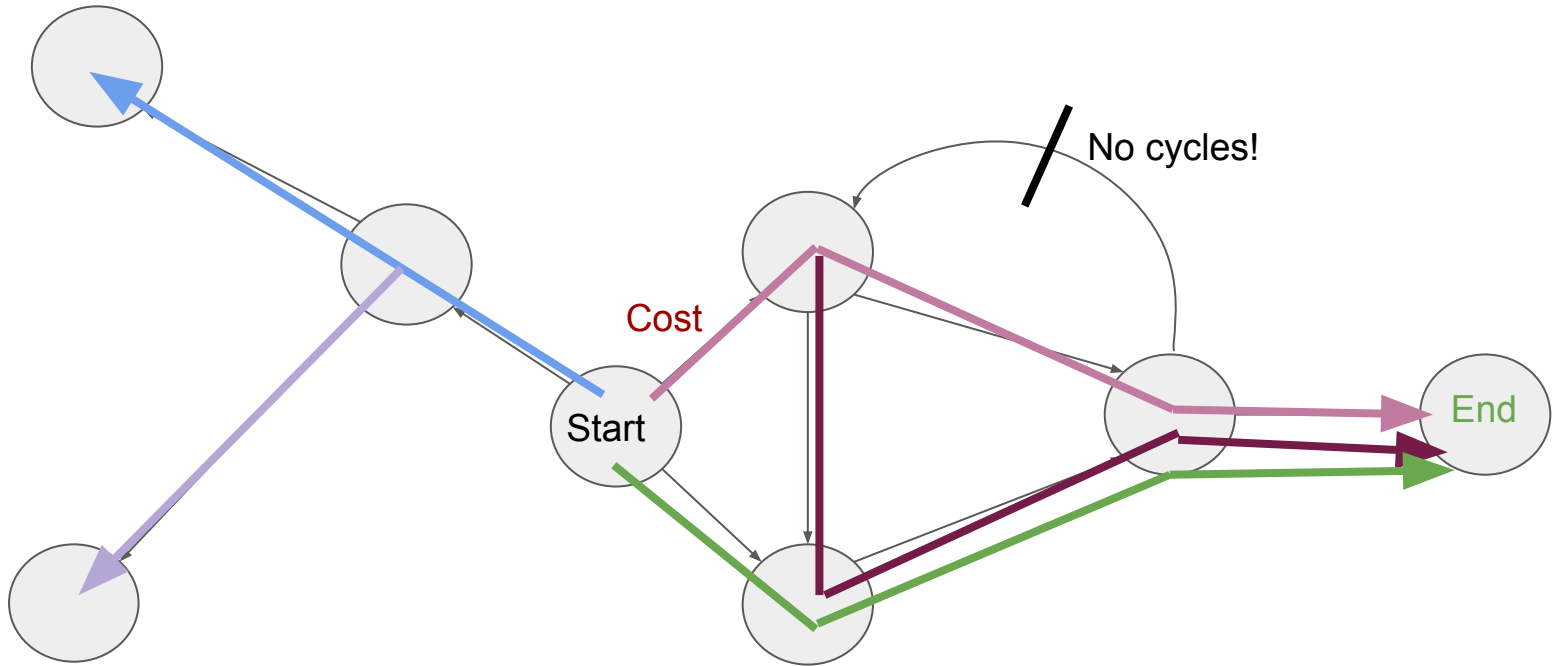
Runtime:  $O(b^d)$

Heavy in space:  $O(b^d)$

# Remember subproblems! DP



# Remember subproblems! DP

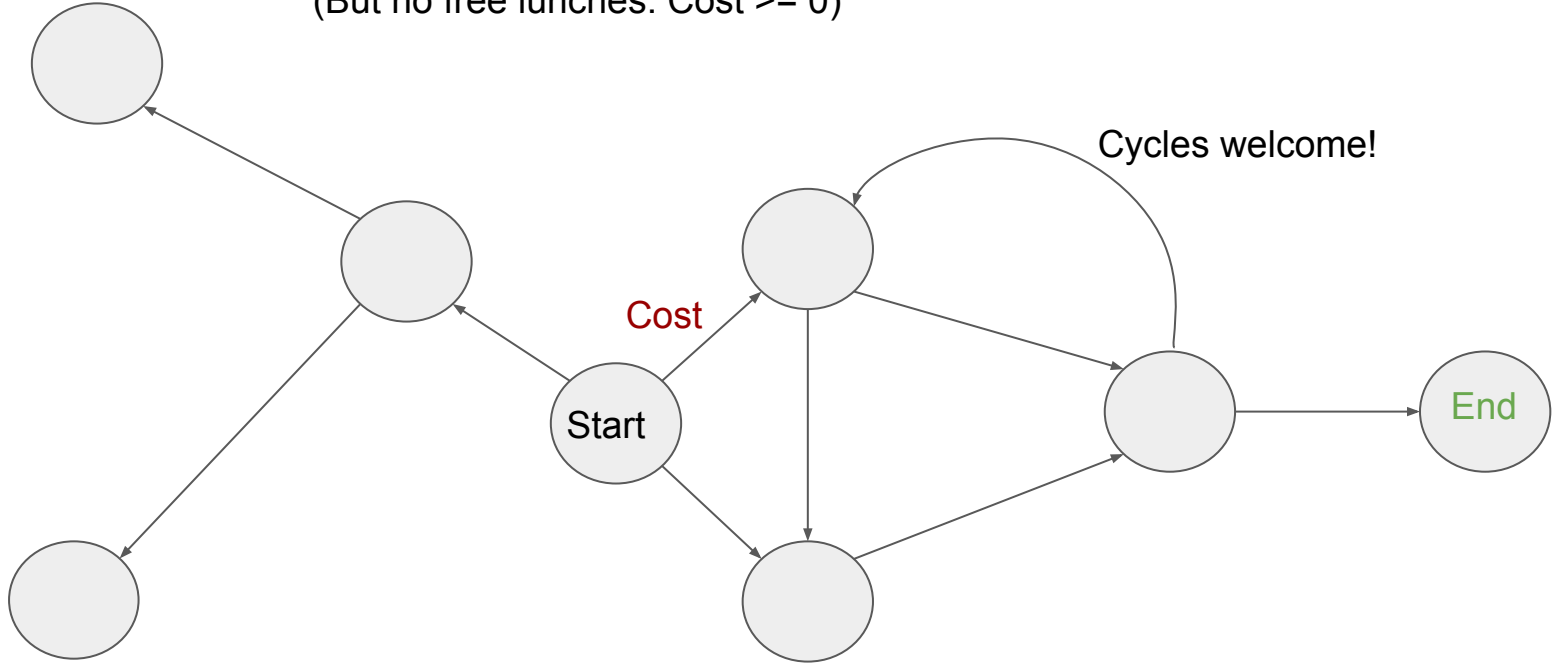


Runtime:  $O(\text{\#states} * b)$

Memory:  $O(\text{\#states})$

# Path of least resistance with memory! UCS

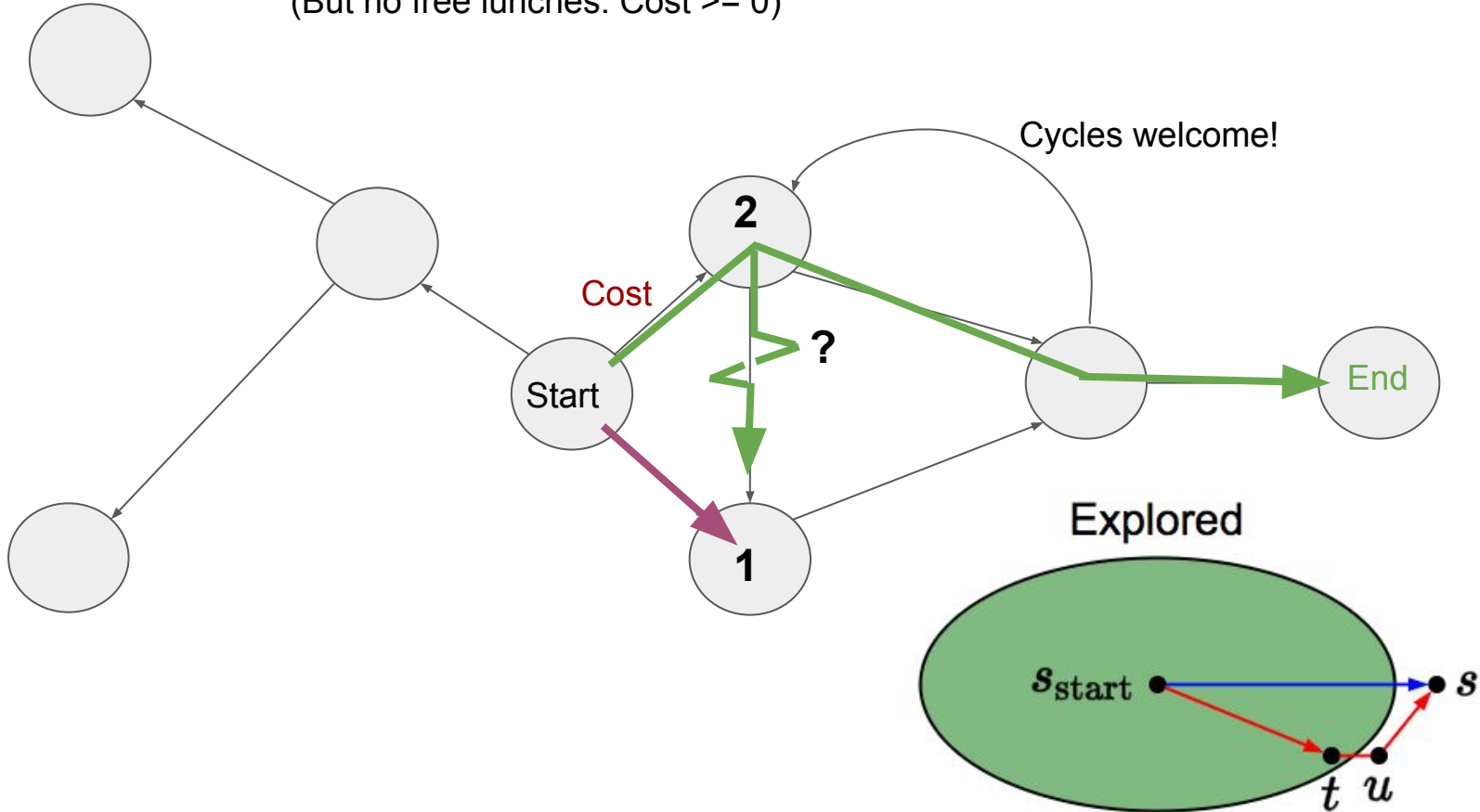
(But no free lunches. Cost  $\geq 0$ )





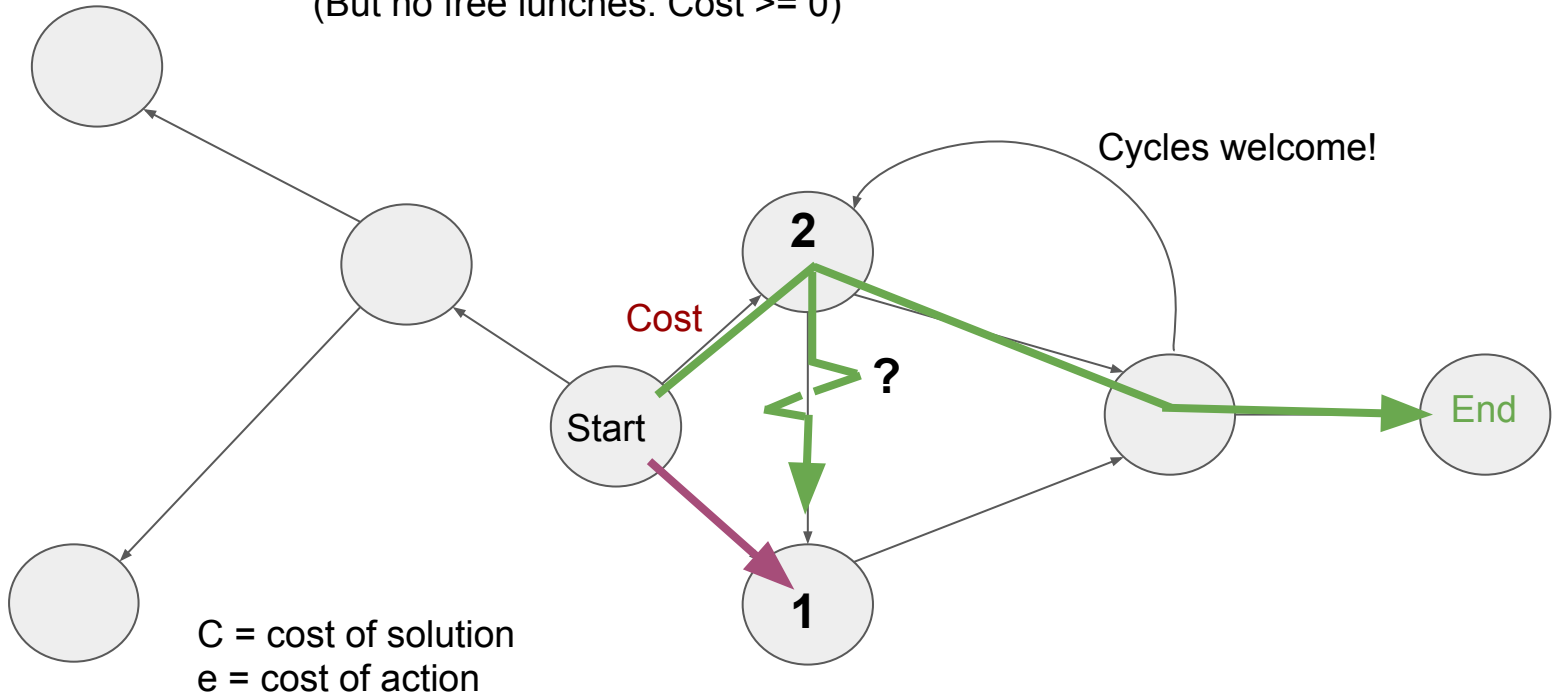
# Path of least resistance with memory! UCS

(But no free lunches. Cost  $\geq 0$ )



# Path of least resistance with memory! UCS

(But no free lunches. Cost  $\geq 0$ )



Runtime:  $\sim O(\min(b^{(C/e)}, \text{\#states} * b))$       Memory = Runtime

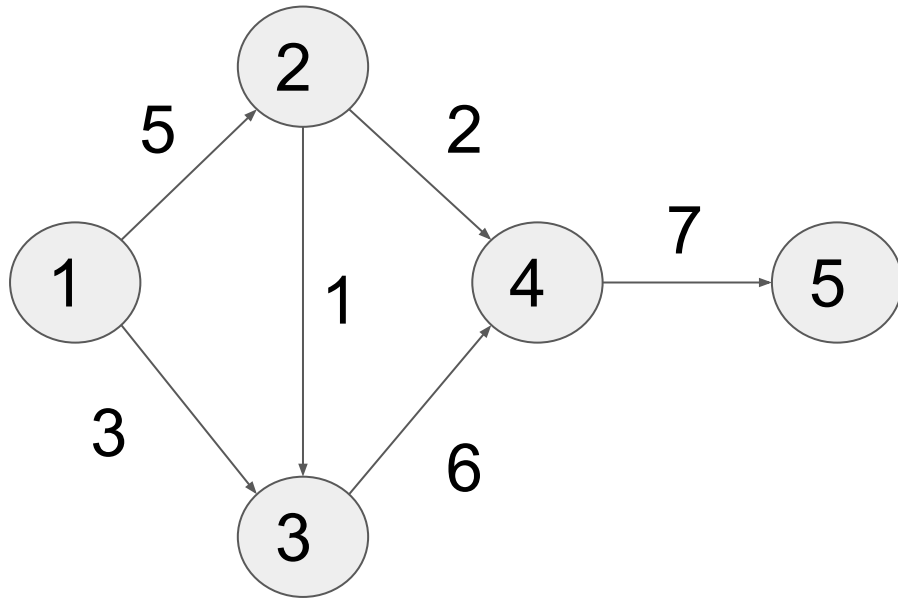
# Problem

There exists  $N$  cities, conveniently labelled from 1 to  $N$ .

There are roads connecting some pairs of cities. The road connecting city  $i$  and city  $j$  takes  $c(i,j)$  time to traverse. However, one can only travel from a city with smaller label to a city with larger label (i.e. each road is one-directional).

From city 1, we want to travel to city  $N$ . What is the shortest time required to make this trip, given the additional constraint that we should visit more odd-labeled cities than even labeled cities?

# Example



Best path is [1, 3, 4, 5] with cost 16.

[1, 2, 4, 5] has cost 14 but visits equal number of odd and even cities.

# State Representation



**Key idea: state**

A **state** is a summary of all the past actions sufficient to choose future actions **optimally**.

# State Representation

We need to know where we are currently at: **current\_city**

We need to know how many odd and even cities we have visited thus far: **#odd, #even**

State Representation: (**current\_city, #odd, #even**)

Total number of states:  $O(N^3)$

## Can We Do Better?

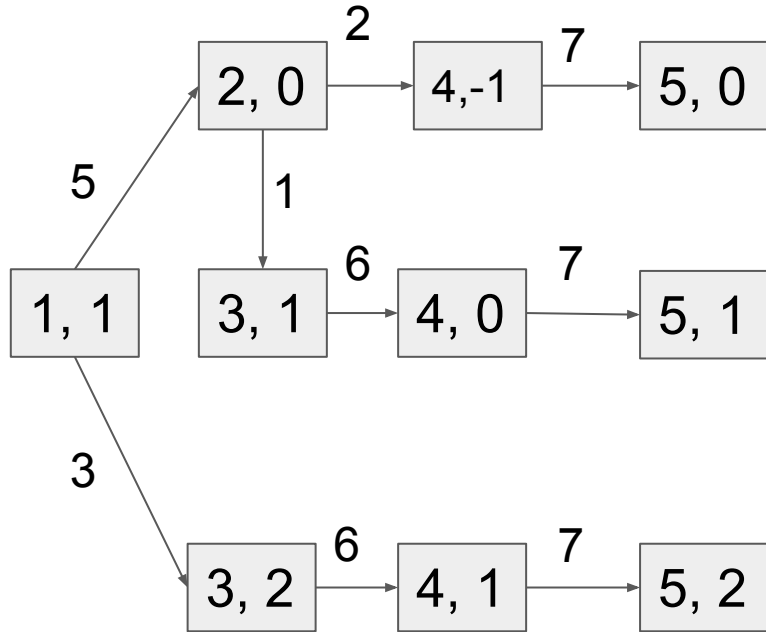
Check if all the information is really required

We store **#odd** and **#even** so that we can check whether **#odd - #even > 0** at **(N, #odd, #even)**

Why not store **#odd - #even** directly instead?

**(current\_city, #odd - #even)** --  $O(N^2)$  states

# State Graph





# Precise Formulation of Problem

Let  $E$  be the set of roads between cities

State  $s = (i, d)$  -- (current\_city, #odd - #even)

Actions( $s$ ): {move( $j$ ) for  $(i, j) \in E$ }

Cost( $s$ , move( $j$ )):  $c(i, j)$

Succ( $s$ ,  $a$ ):  $(j, d + 1)$  if  $j$  is odd,  $(j, d - 1)$  otherwise

Start:  $(1, 1)$

isGoal( $s$ ):  $i = N$  and  $d > 0$

# Solving the Problem

Since we are computing shortest path, which is some form of optimization, we consider DP and UCS.

Recall

- DP can handle negative edges but works only on DAGs
- UCS works on general graphs, but cannot handle negative edges

Since we have a DAG and all edges are positive, both work!

# Solving the Problem: Uniform Cost Search



**Algorithm: uniform cost search [Dijkstra, 1956]**

Add  $s_{\text{start}}$  to **frontier** (priority queue)

Repeat until frontier is empty:

    Remove  $s$  with smallest priority  $p$  from frontier

    If  $\text{IsGoal}(s)$ : return solution

    Add  $s$  to **explored**

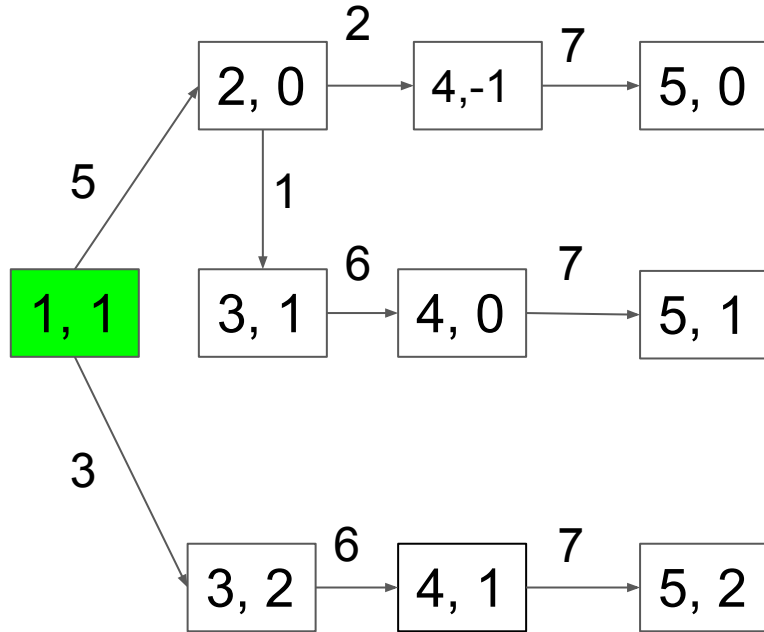
    For each action  $a \in \text{Actions}(s)$ :

        Get successor  $s' \leftarrow \text{Succ}(s, a)$

        If  $s'$  already in explored: continue

        Update **frontier** with  $s'$  and priority  $p + \text{Cost}(s, a)$

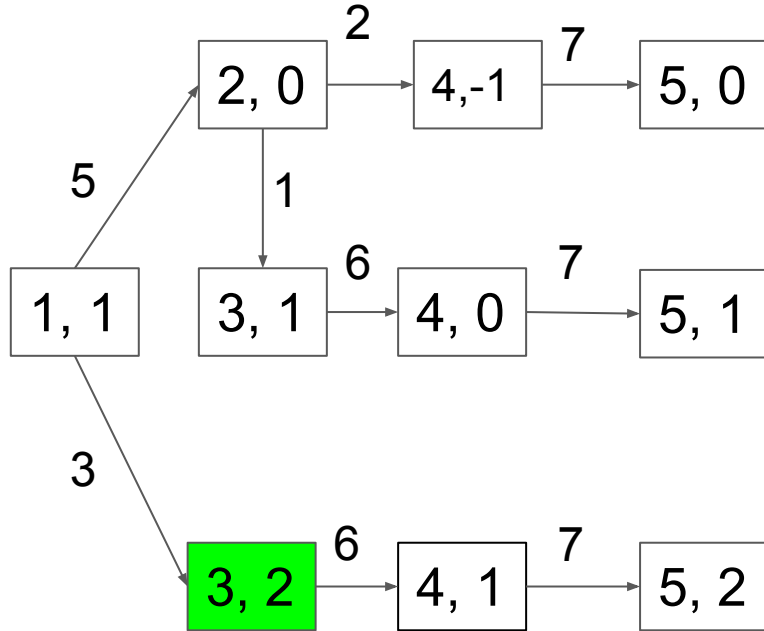
# Simulation of UCS



Explored:  
(1, 1) : 0

Frontier:  
(2, 0) : 5  
(3, 2) : 3

# Simulation of UCS



Explored:

(1, 1) : 0

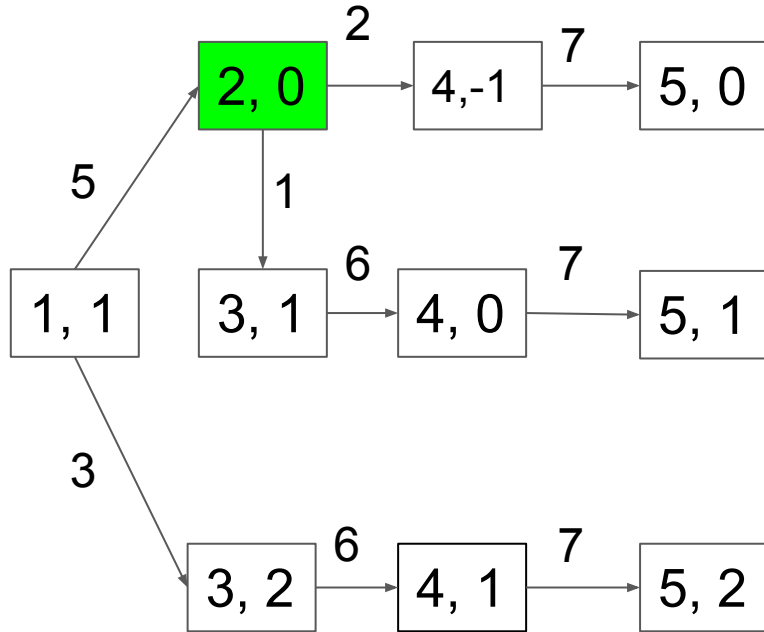
(3, 2) : 3

Frontier:

(2, 0) : 5

(4, 1) : 9

# Simulation of UCS



Explored:

$(1, 1) : 0$

$(3, 2) : 3$

$(2, 0) : 5$

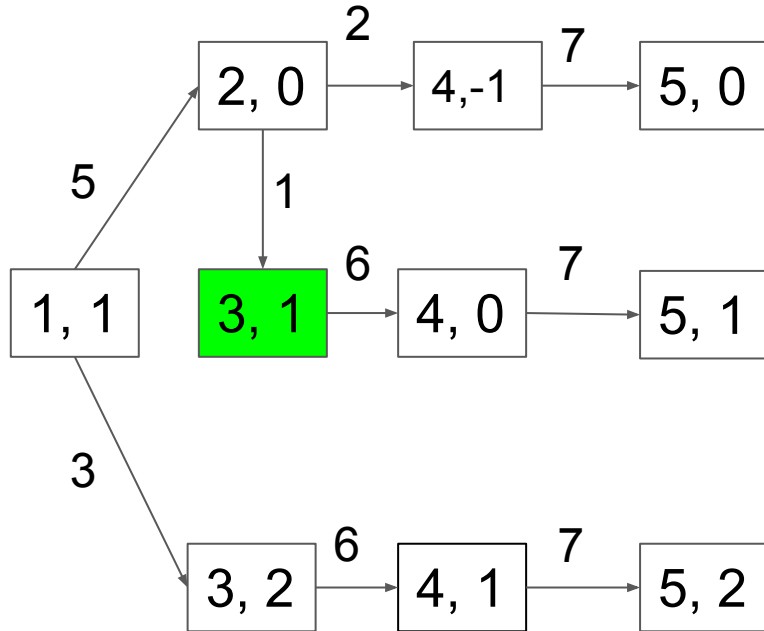
Frontier:

$(4, 1) : 9$

$(4, -1) : 7$

$(3, 1) : 6$

# Simulation of UCS



Explored:

(1, 1) : 0

(3, 2) : 3

(2, 0) : 5

(3, 1) : 6

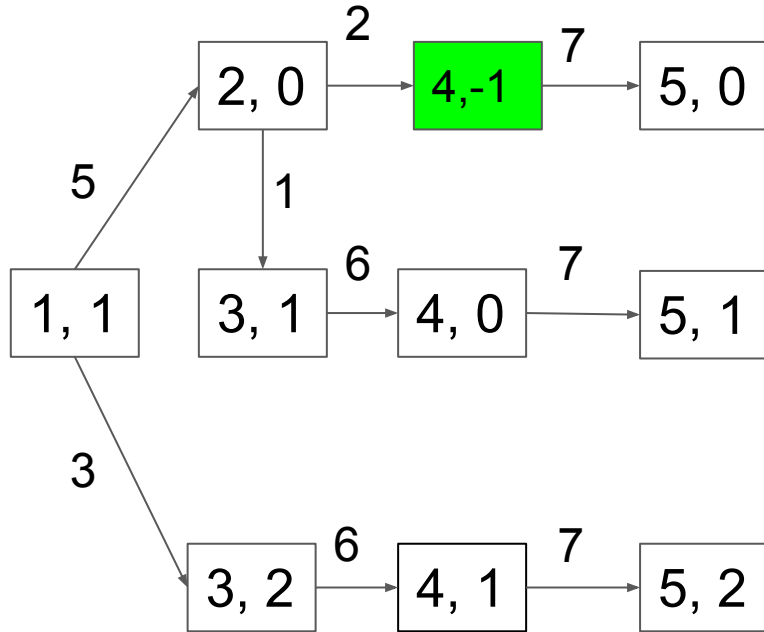
Frontier:

(4, 1) : 9

(4, -1) : 7

(4, 0) : 12

# Simulation of UCS



Explored:

(1, 1) : 0

(3, 2) : 3

(2, 0) : 5

(3, 1) : 6

(4, -1) : 7

Frontier:

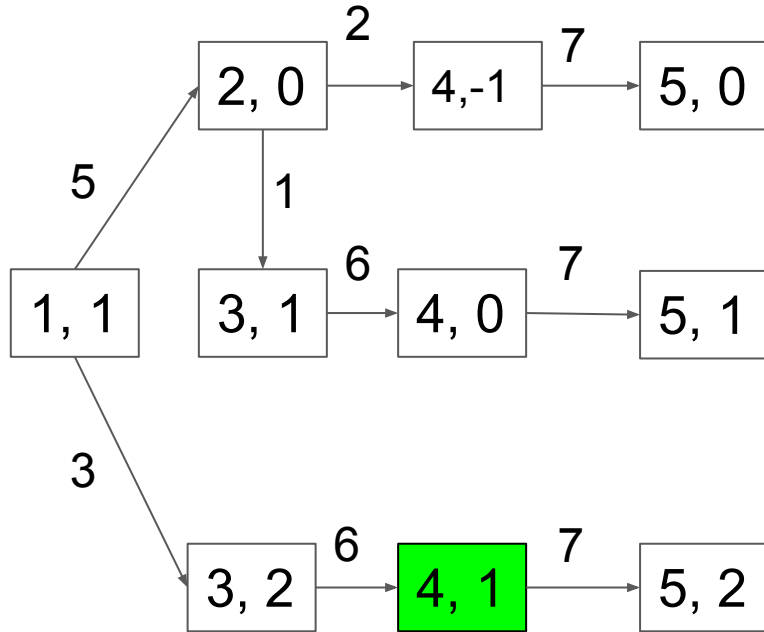
(4, 1) : 9

(4, 0) : 12

(5, 0) : 14



# Simulation of UCS



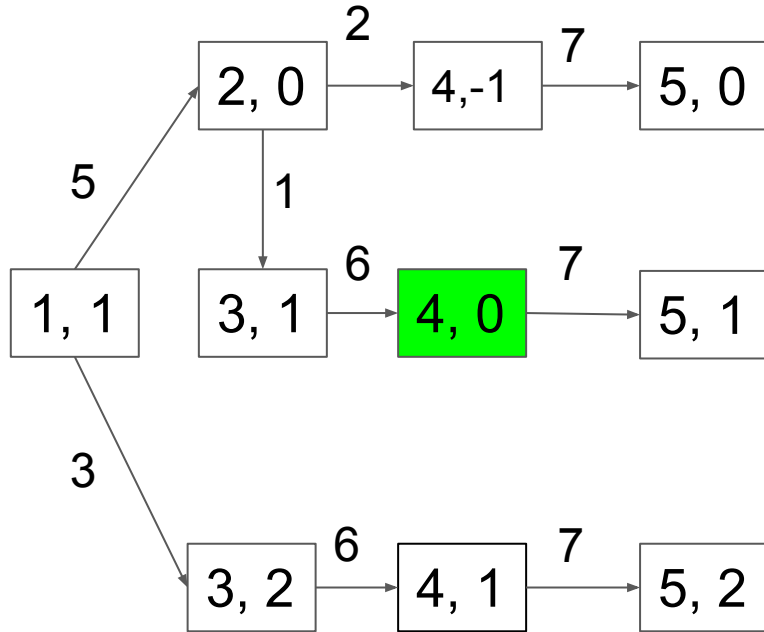
Explored:

(1, 1) : 0  
(3, 2) : 3  
(2, 0) : 5  
(3, 1) : 6  
(4, -1) : 7  
(4, 1) : 9

Frontier:

(4, 0) : 12  
(5, 0) : 14  
(5, 2) : 16

# Simulation of UCS



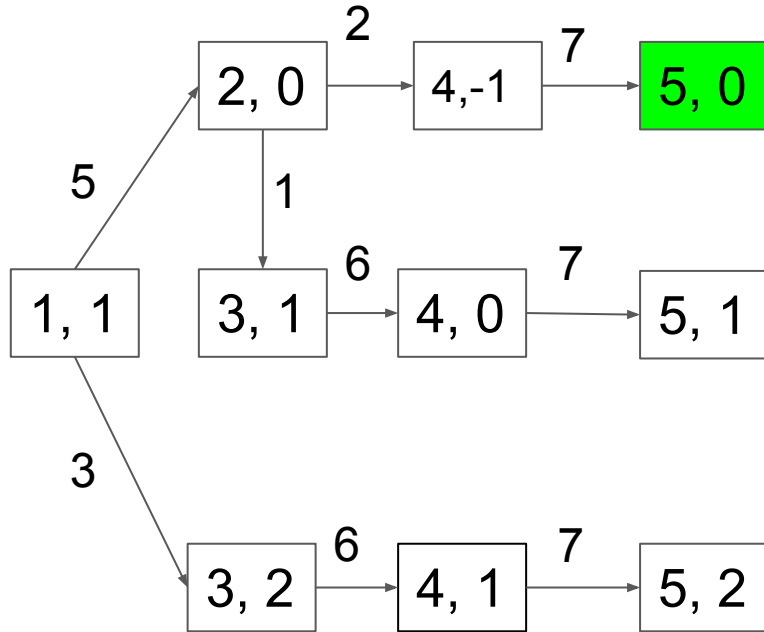
Explored:

(1, 1) : 0  
(3, 2) : 3  
(2, 0) : 5  
(3, 1) : 6  
(4, -1) : 7  
(4, 1) : 9  
(4, 0) : 12

Frontier:

(5, 0) : 14  
(5, 2) : 16  
(5, 1) : 19

# Simulation of UCS



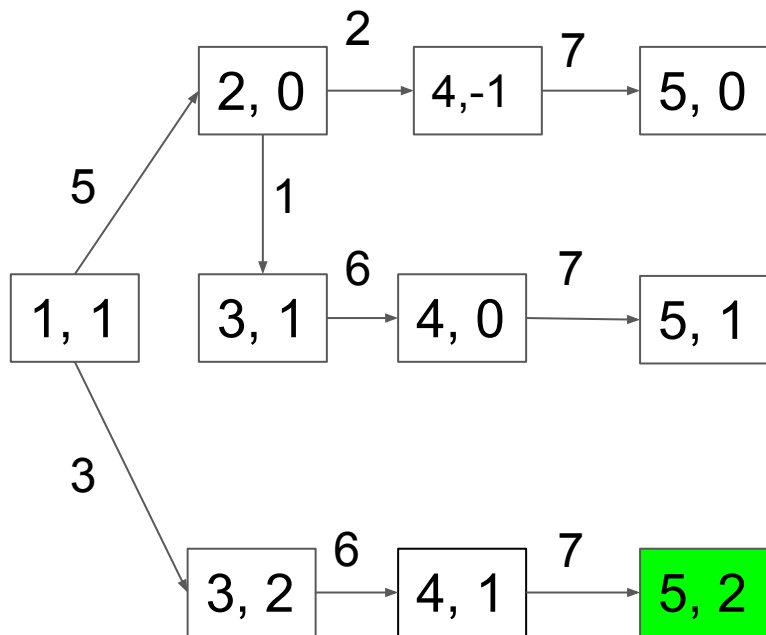
Explored:

(1, 1) : 0  
(3, 2) : 3  
(2, 0) : 5  
(3, 1) : 6  
(4, -1) : 7  
(4, 1) : 9  
(4, 0) : 12  
(5, 0) : 14

Frontier:

(5, 2) : 16  
(5, 1) : 19

# Simulation of UCS



Explored:

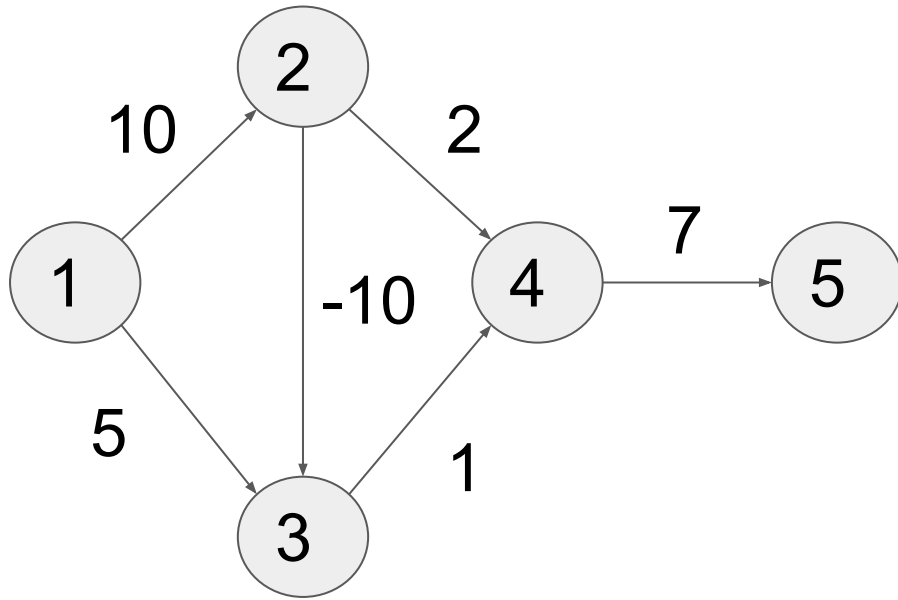
(1, 1) : 0  
(3, 2) : 3  
(2, 0) : 5  
(3, 1) : 6  
(4, -1) : 7  
(4, 1) : 9  
(4, 0) : 12  
(5, 0) : 14  
(5, 2) : 16

Frontier:

(5, 1) : 19

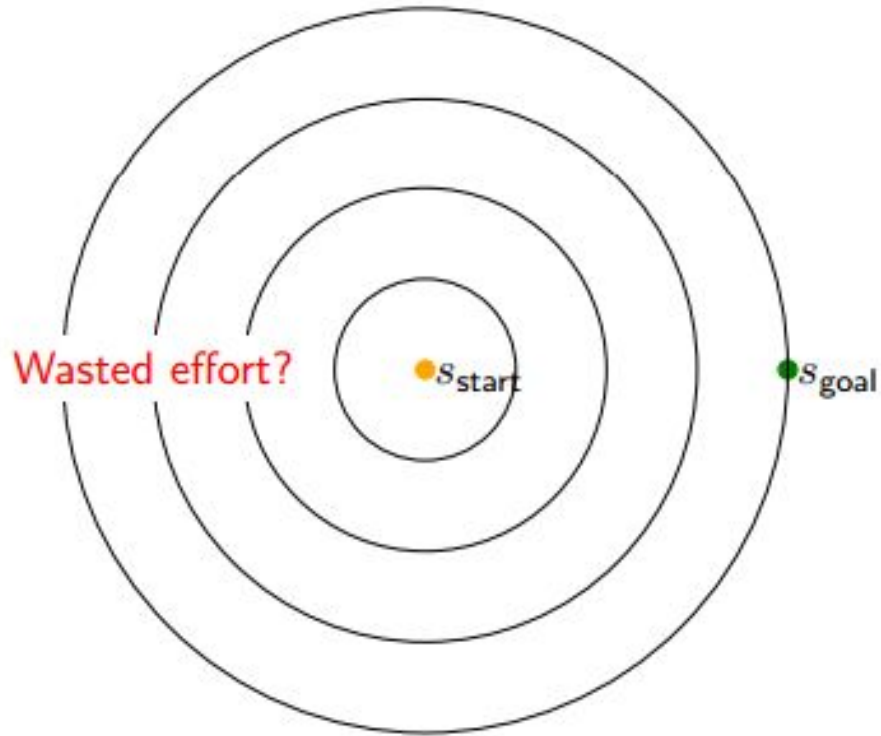
**STOP!**

# UCS cannot handle negative edge weights



Best path is [1,2,3,4,5] with cost of 8, but UCS will output [1,3,4,5] with cost of 13 because 3 is set as 'explored' before 2.

# Improve UCS: A\* Search



# Recap of A\* Search

- Modify the cost of edges and run UCS on the new graph
  - $\text{Cost}'(s, a) = \text{Cost}(s, a) + h(\text{Succ}(s, a)) - h(s)$
- $h(s)$  is a heuristic that is our estimate of  $\text{FutureCost}(s)$
- If  $h(s)$  is consistent then the modified edge weights will return min cost path
- One can find a good consistent  $h$  by performing relaxation
- If  $c$  is min cost on original graph,  $c'$  is min cost on modified graph, then  $c' = c + h(s_{\text{goal}}) - h(s_{\text{start}})$

# Relaxation

A good way to come up with a reasonable heuristic is to solve an easier (less constrained) version of the problem

For example, we can remove the constraint that we visit more odd cities than even cities.

$h(s) = h((i, d)) = \text{length of shortest path from city } i \text{ to city } N$



## Note on Relaxation

The main point of relaxation is to attain a problem that **can be solved more efficiently**.

In our case, the modified shortest path problem has  $O(N)$  states instead of  $O(N^2)$  can thus can be solved more efficiently

# Checking consistency

- $\text{Cost}(s, a) + h(\text{Succ}(s, a)) - h(s) \geq 0$  (Triangle Inequality)
  - Suppose  $s = (i, d)$  and  $\text{Succ}(s, a) = (j, d')$
  - Note that  $h((i, d)) - h((j, d')) \leq c(i, j) = \text{Cost}(s, a)$
- $h((N, d)) = 0$

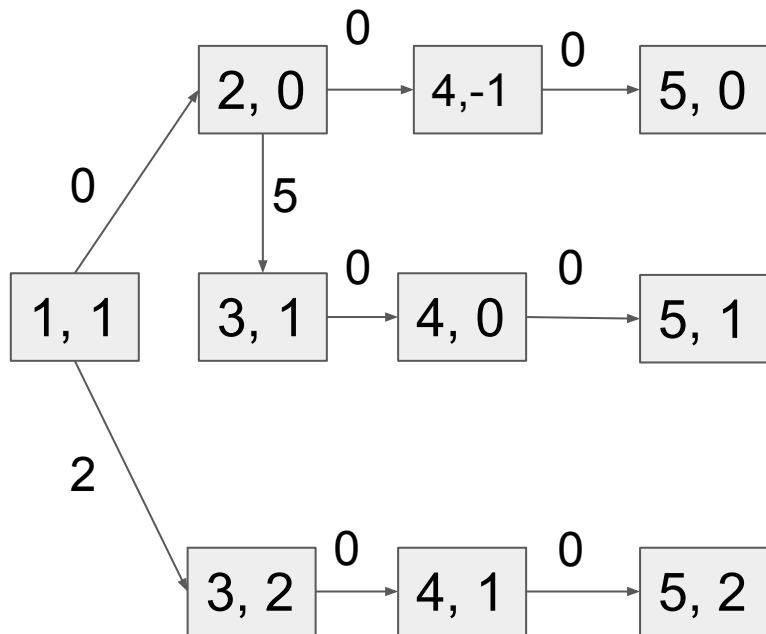
# How to compute h?

We can reverse the direction of all edges, and then perform UCS starting from city N, and our goal state is city 1.

This takes  $O(n \log n)$  time, where  $n$  is the number of states whose distance to city N is no farther than the distance of city 1 to city N

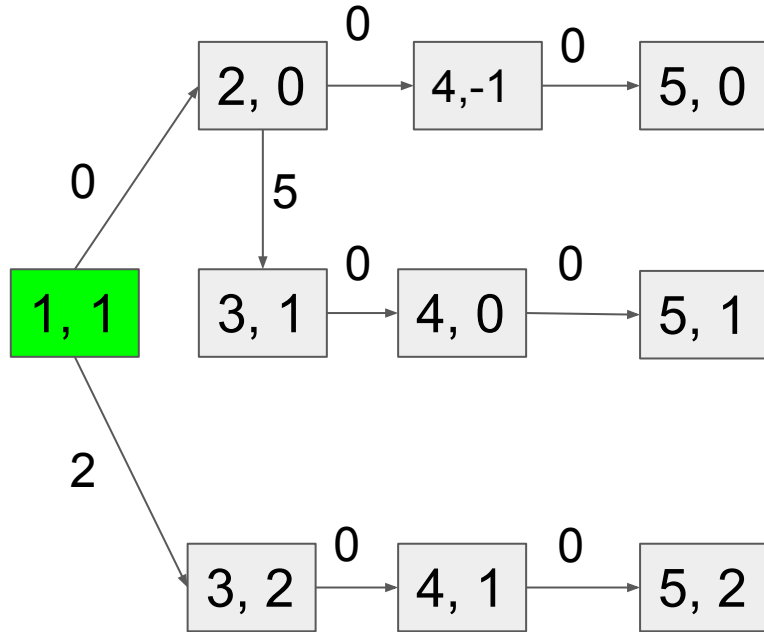
city	1	2	3	4	5
h	14	9	13	7	0

# Modified State Graph



$$\text{Cost}(s, a) + h(\text{Succ}(s, a)) - h(s)$$

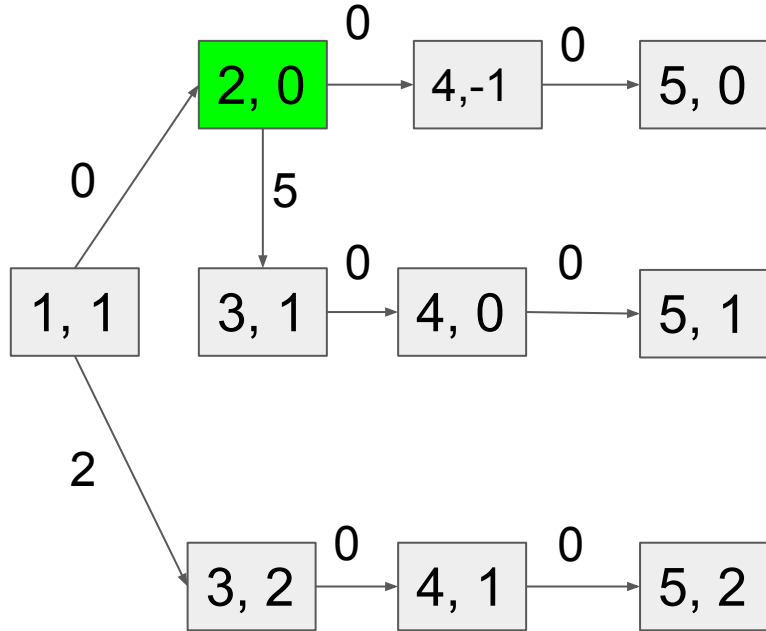
# Simulation of UCS (A\*)



Explored:  
(1, 1) : 0

Frontier:  
(2, 0) : 0  
(3, 2) : 2

# Simulation of UCS (A\*)



Explored:

(1, 1) : 0

(2, 0) : 0

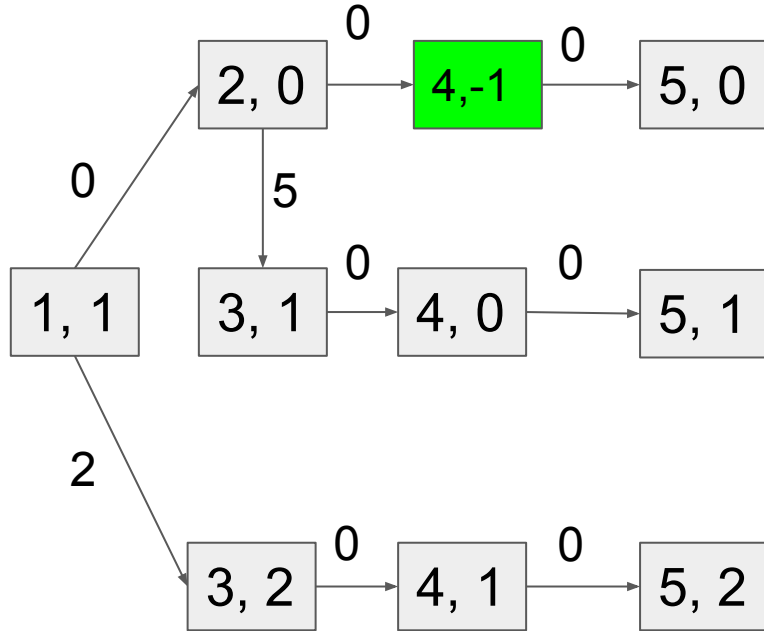
Frontier:

(3, 2) : 2

(3, 1) : 3

(4, -1) : 0

# Simulation of UCS (A\*)



Explored:

(1, 1) : 0

(2, 0) : 0

(4, -1) : 0

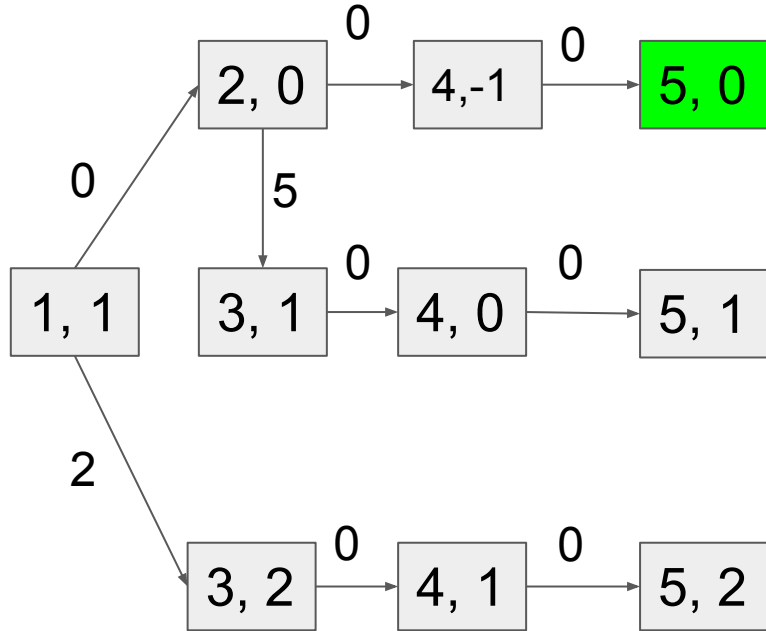
Frontier:

(3, 2) : 2

(3, 1) : 3

(5, 0) : 0

# Simulation of UCS (A\*)



Explored:

(1, 1) : 0

(2, 0) : 0

(4, -1) : 0

(5, 0) : 0

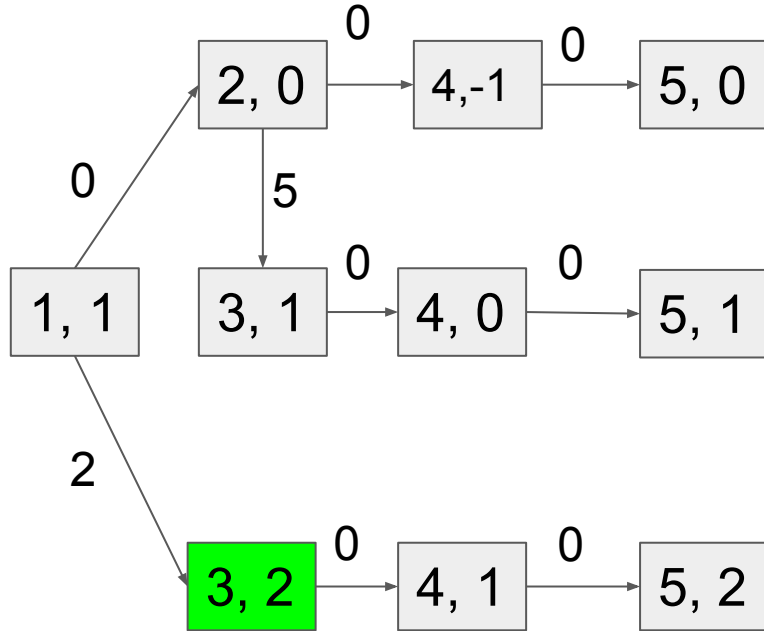
Frontier:

(3, 2) : 2

(3, 1) : 3



# Simulation of UCS (A\*)



Explored:

(1, 1) : 0

(2, 0) : 0

(4, -1) : 0

(5, 0) : 0

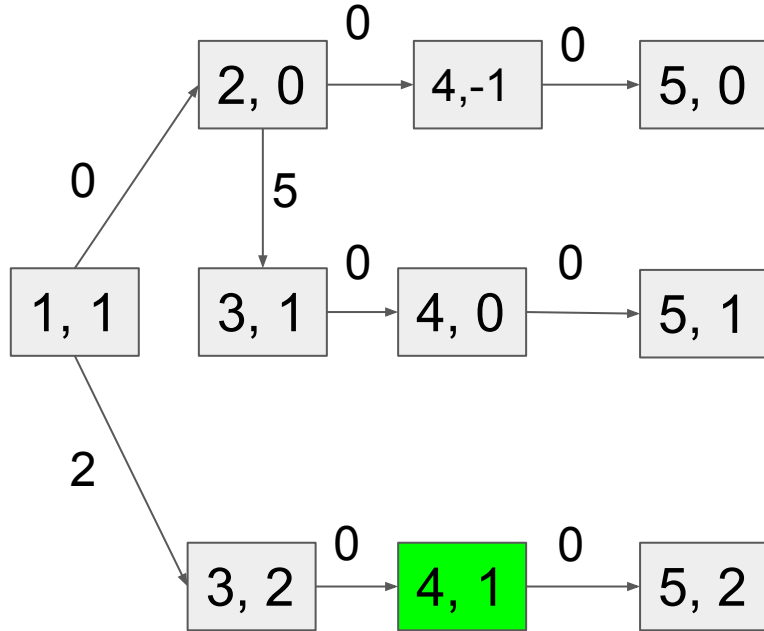
(3, 2) : 2

Frontier:

(3, 1) : 3

(4, 1) : 2

# Simulation of UCS (A\*)



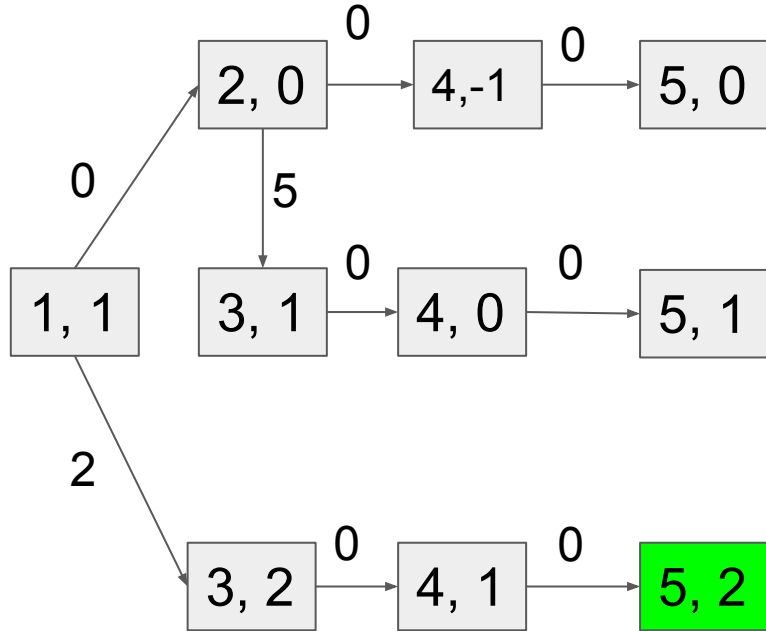
Explored:

(1, 1) : 0  
(2, 0) : 0  
(4, -1) : 0  
(5, 0) : 0  
(3, 2) : 2  
(4, 1) : 2

Frontier:

(3, 1) : 3  
(5, 2) : 0

# Simulation of UCS (A\*)



Explored:

(1, 1) : 0  
(2, 0) : 0  
(4, -1) : 0  
(5, 0) : 0  
(3, 2) : 2  
(4, 1) : 2  
(5, 2) : 2

Frontier:

(3, 1) : 3

**STOP!**

Actual Cost is  $2 + h(1) = 2 + 14 = 16$

# Comparison of States visited

## UCS

Explored:

(1, 1) : 0

(3, 2) : 3

(2, 0) : 5

(3, 1) : 6

(4, -1) : 7

(4, 1) : 9

(4, 0) : 12

(5, 0) : 14

(5, 2) : 16

Frontier:

(5, 1) : 19

## UCS(A\*)

Explored:

(1, 1) : 0

(2, 0) : 0

(4, -1) : 0

(5, 0) : 0

(3, 2) : 2

(4, 1) : 2

(5, 2) : 2

Frontier:

(3, 1) : 3

# Summary

- States Representation/Modelling
  - make state representation as compact as possible, remove unnecessary information
- DP
  - underlying graph cannot have cycles
  - visit all reachable states
- UCS
  - actions cannot have negative cost
  - visit only a subset of states
- $A^*$ 
  - ensure that relaxed problem can be solved more efficiently