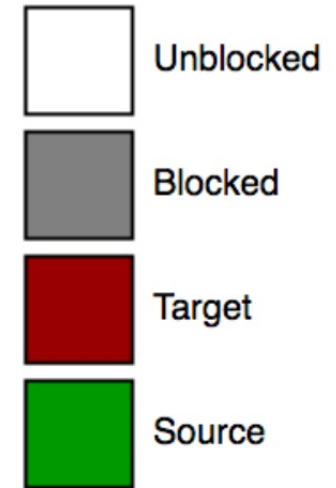
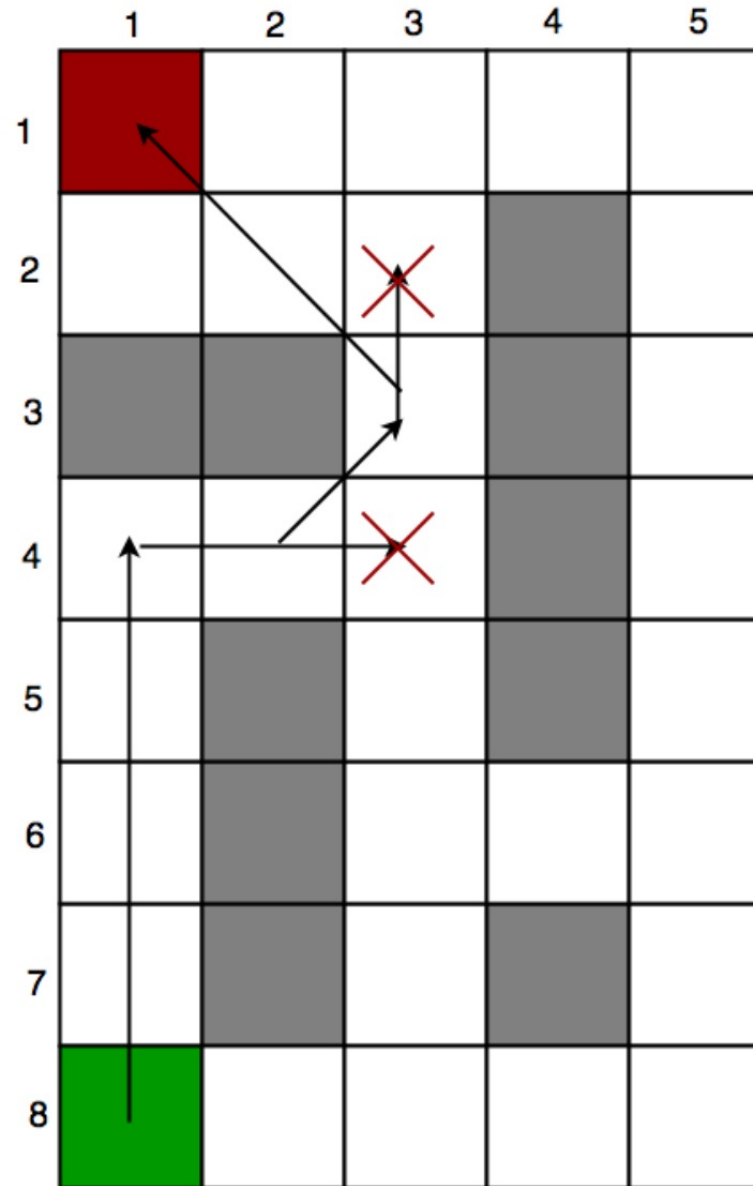


CSCI 3202: Intro to Artificial Intelligence

UCS, A* Search and Heuristics

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A* Search Algorithm makes the most intelligent choice at each step. Hence you can see that algorithm goes from (4,2) to (3,3) and not (4,3) (shown by cross).

Similarly the algorithm goes from (3,3) to (2,2) and not (2,3) (shown by cross).

[Source](#)

Uniform-cost Search (UCS)

- Expand out in contours, where least cost dictates which nodes we explore.
- Eventually, we will find a path to the goal - but the search is not directed
- BFS strategy
- Expand cheapest node first (lowest path cost)
- Frontier is a priority queue
- Cost function sets priority

Uniform-cost Search (UCS) – pseudocode

function UNIFORM-COST-SEARCH(*problem*) **returns** a solution, or failure

node \leftarrow a node with STATE = *problem*.INITIAL-STATE, PATH-COST = 0

frontier \leftarrow a priority queue ordered by PATH-COST, with *node* as the only element

explored \leftarrow an empty set

loop do

if EMPTY?(*frontier*) **then return** failure

node \leftarrow POP(*frontier*) /* chooses the lowest-cost node in *frontier* */

if *problem*.GOAL-TEST(*node*.STATE) **then return** SOLUTION(*node*)

 add *node*.STATE to *explored*

for each *action* **in** *problem*.ACTIONS(*node*.STATE) **do**

child \leftarrow CHILD-NODE(*problem*, *node*, *action*)

if *child*.STATE is not in *explored* or *frontier* **then**

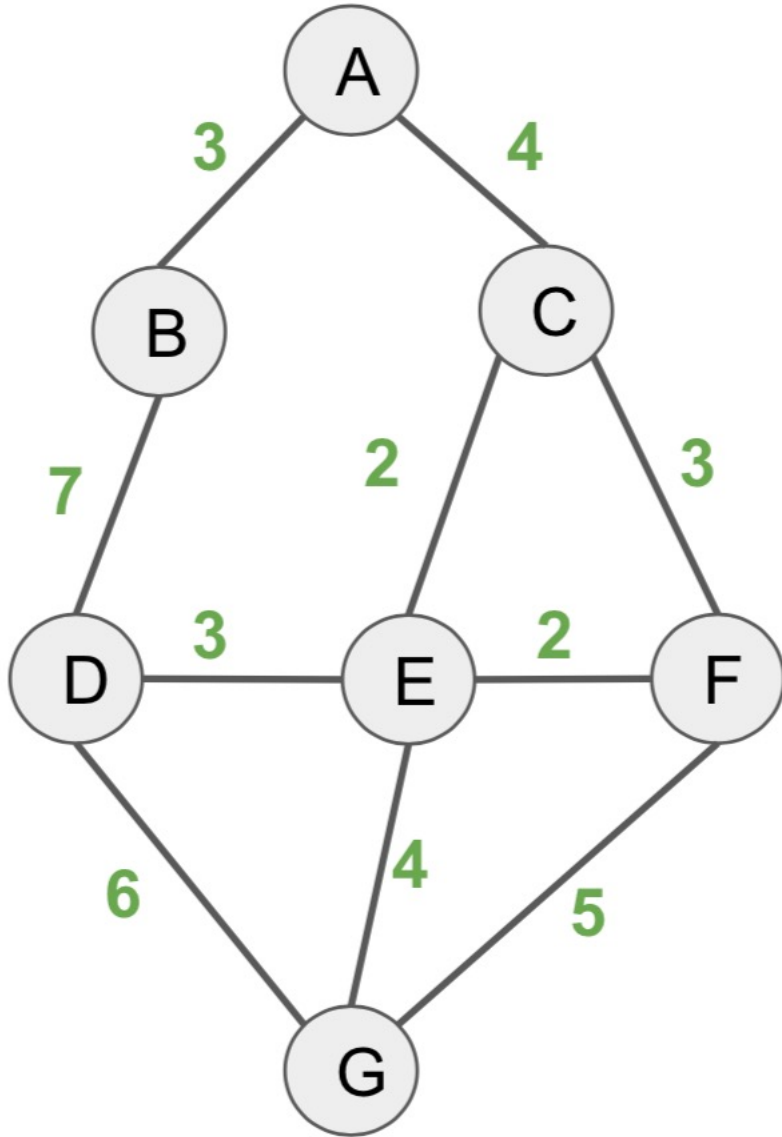
frontier \leftarrow INSERT(*child*, *frontier*)

else if *child*.STATE is in *frontier* with higher PATH-COST **then**

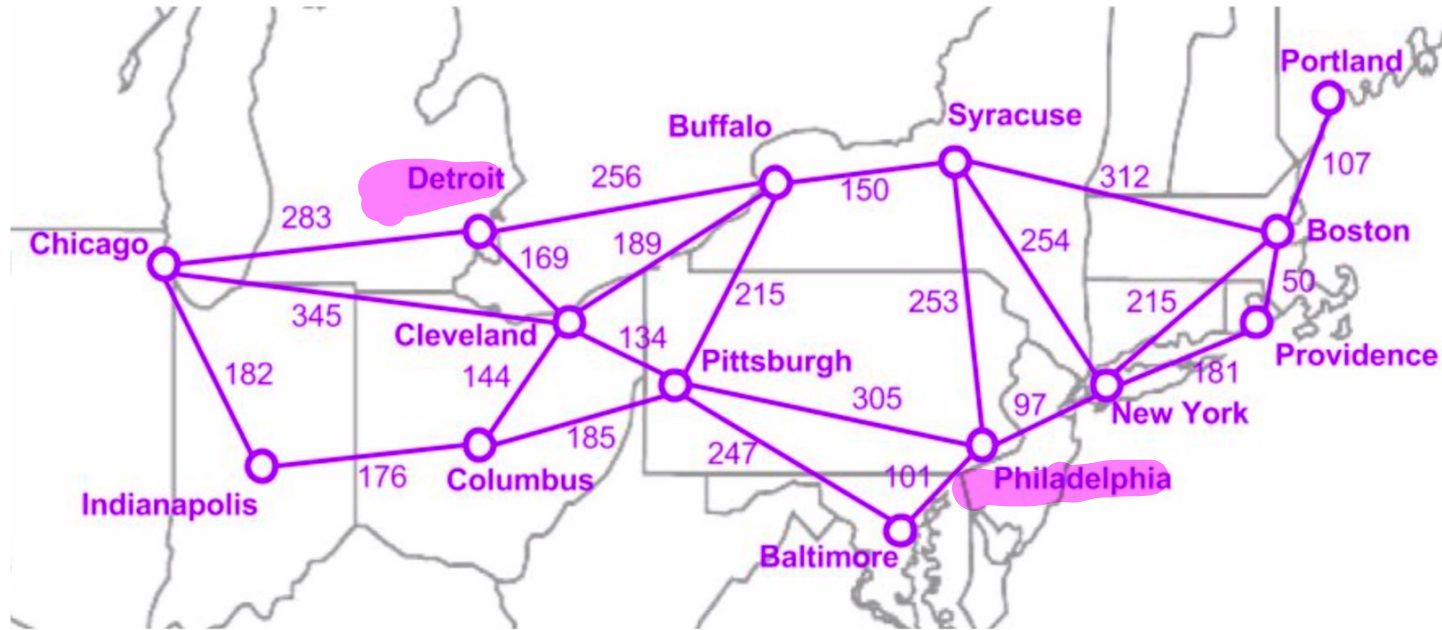
 replace that *frontier* node with *child*

Uniform-cost Search (UCS)

Example: Perform a UCS on the graph below. A is the starting point; G is the goal.



Uniform-cost Search (UCS)



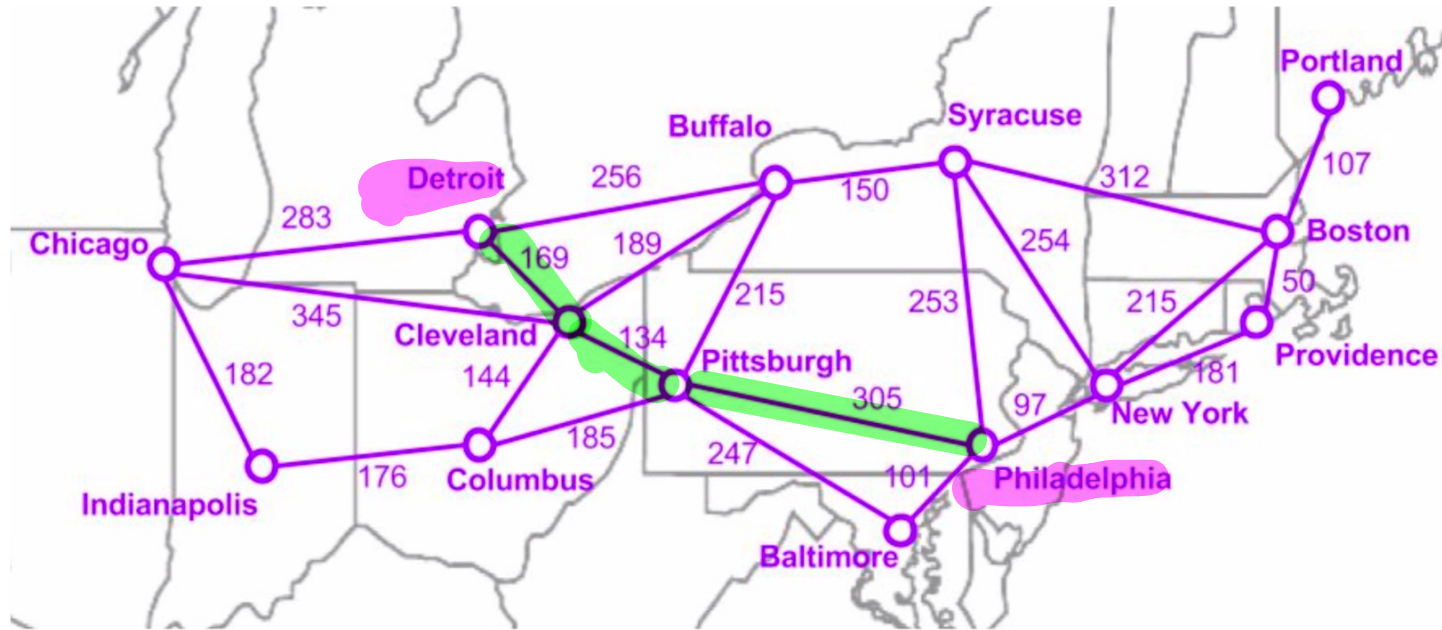
Example: Use UCS to find a route from Detroit to Philadelphia.

6. $Ch(514)$, Re ,
 $Ph(608)$, cl ,
 $Ba(550)$, Bu ,
 $Bo(718)$, Pi ,
 $Ny(660)$, Sy

$Sy \rightarrow Ny, Bu, Ph$
 $Ny = 406 + 254$
 $Bo = 406 + 312$
 $Ph = 406 + 256 = 662$

<u>F</u>	<u>E</u>	<u>ACT</u>
1. $De(0)$		
2. $Bu(256)$ $cl(169)$ $ch(283)$	De	$De \rightarrow Bu, Cl, Ch$
3. $Bu(256)$ $Pi(303)$ $ch(514)$	$De,$ cl	$cl \rightarrow Bu, Pi, Ch$ $Bu = 169 + 189$ $Pi = 169 + 134$ $Ch = 169 + 345$
4. $Sy(406)$ $Pi(303)$ $ch(514)$	$De,$ $cl,$ Bu	$Bu \rightarrow Sy, Pi,$ cl $Sy = 256 + 150$ $Pi = 256 + 215$
5. $Sy(406)$ $ch(514)$ $Ph(608)$ $Ba(550)$	De $cl,$ $Bu,$ Pi	$Pi \rightarrow Ph$ Ba, Bu $Ph = 303 + 305$ $Ba = 303 + 247$

Uniform-cost Search (UCS)



Example: Use UCS to find a route from Detroit to Philadelphia.

6. $Ch(514)$, Re , $Sy \rightarrow Ny, Bo, Ph$
 $Ph(608)$, cl , $Ny = 406 + 254$
 $Ba(550)$, Bu , $Bo = 406 + 312$
 $Bo(718)$, Pi , $Ph = 406 + 256 = 662$
 $Ny(660)$, Sy

Det \rightarrow Clu, Pit, Ph = $169 + 134 + 305 = 608$

F	E	ACT
7. $ln(696)$ $Ph(608)$ $Ba(550)$ $Bo(718)$ $Ny(660)$	De , cl , Bu Pi Sy Ch	$Ch \rightarrow \cancel{De}, \cancel{cl}, ln$ $ln \rightarrow 514 + 182$
8. $ln(696)$ $Ph(608)$ $Bo(718)$ $Ny(660)$	De , cl , Bu , Pi , Sy Ch , Ba	$Ba \rightarrow Ph, \cancel{Pi}$, $Ph = 550 + 101$ 651 X
9. $Ph \rightarrow 608$		

Uniform-cost Search (UCS)

- *Goal test occurs when node is selected for expansion*
- Because we know we've taken the cheapest path to get there, UCS is **optimal** if all **edge weights > 0**
- It is also **complete** because it's a more general form of BFS (which is complete)



Uniform-cost Search (UCS)

- Can get stuck if there are sequences of no-cost actions. Optimality requires positive edge weights

$$O(b^{1+\lceil C^*/\epsilon \rceil})$$

- Worst-case in time and space complexity:
 - C^* is cost of optimal solution
 - ϵ is minimal action cost
- Potential inefficiency: Explores in every “direction”

Slightly more informed search – Greedy best-first search

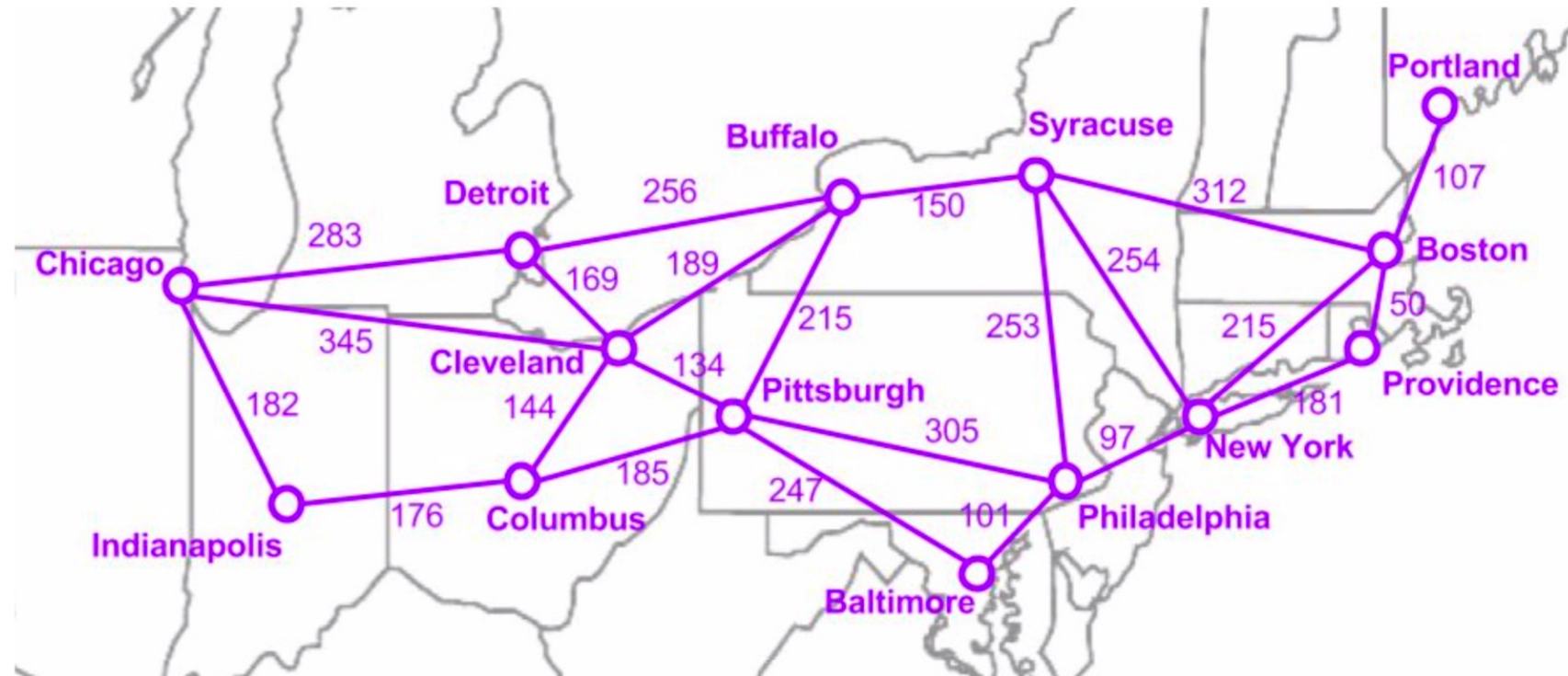
❖ First expand the path that's closest to the goal.

To determine what's closest to the goal, we need to define a heuristic function.

Example: For the traveling in the northeast problem, let's estimate the distance to the goal as the straight-line distance between city and the goal city.

FOR YOUR NEXT STEP,
PICK THE NODE THAT
IS CLOSEST TO YOUR
GOAL

Step costs: miles between cities along major highways



Greedy best-first search

Example: Use the greedy best-first search to find a route from Chicago to Providence.

Distances:

Chicago->833

Detroit->597

Indianapolis->783

Columbus->618

Cleveland->530

Pittsburgh->456

Buffalo->388

Syracuse->256

Baltimore->324

Philadelphia->235

New York->155

Boston->41

Portland->140

$CH \rightarrow DE, CL, IN$

$DE = 597$

$CL = 530$

$IN = 783$

$CL \rightarrow BU, PI, CO$

$BU = 388$

$PI = 456$

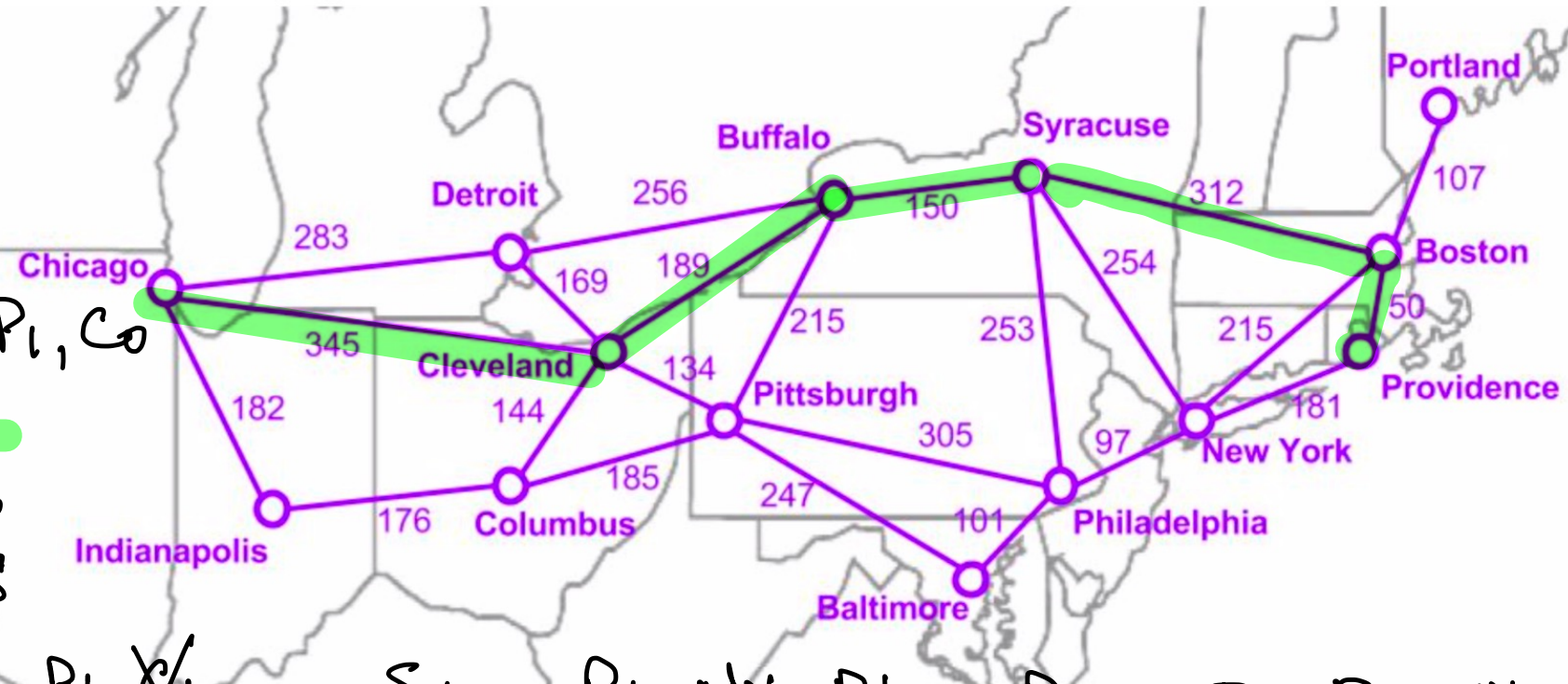
$CO = 618$

$BU \rightarrow SY, PI, \cancel{CL}$

$SY = 256$

$PI = 456$

Heuristic: $h(n)$ = straight-line distance to Providence



$SY \rightarrow BO, NY, PH$

$BO = 41$

$NY = 155$

$PH = 235$

$BO \rightarrow PR, PO, NY$

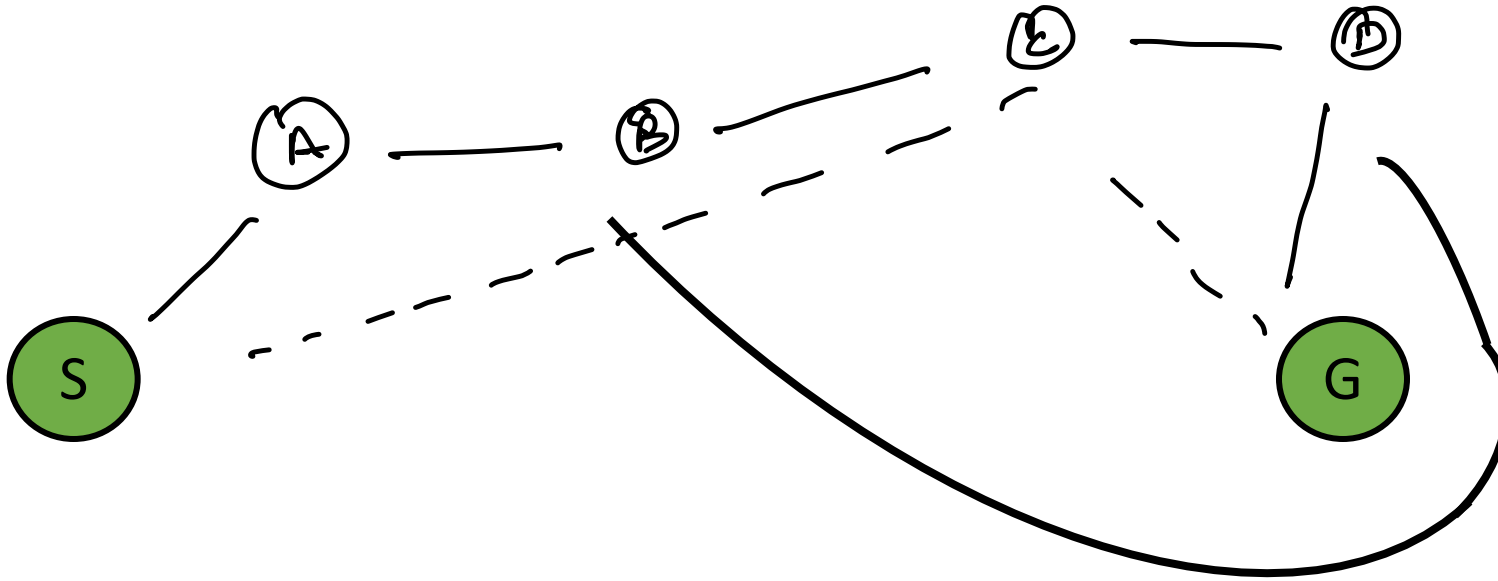
$PR = 0 \checkmark$

$PO = 140$

$NY = 155$

Greedy best-first search

Possible Issue: Won't necessarily find the optimal path. Can get stuck in local optimum.



A* Search

Uniform-cost search:

$$f(n) = g(n) \quad (\text{cost to get to } n)$$

Greedy:

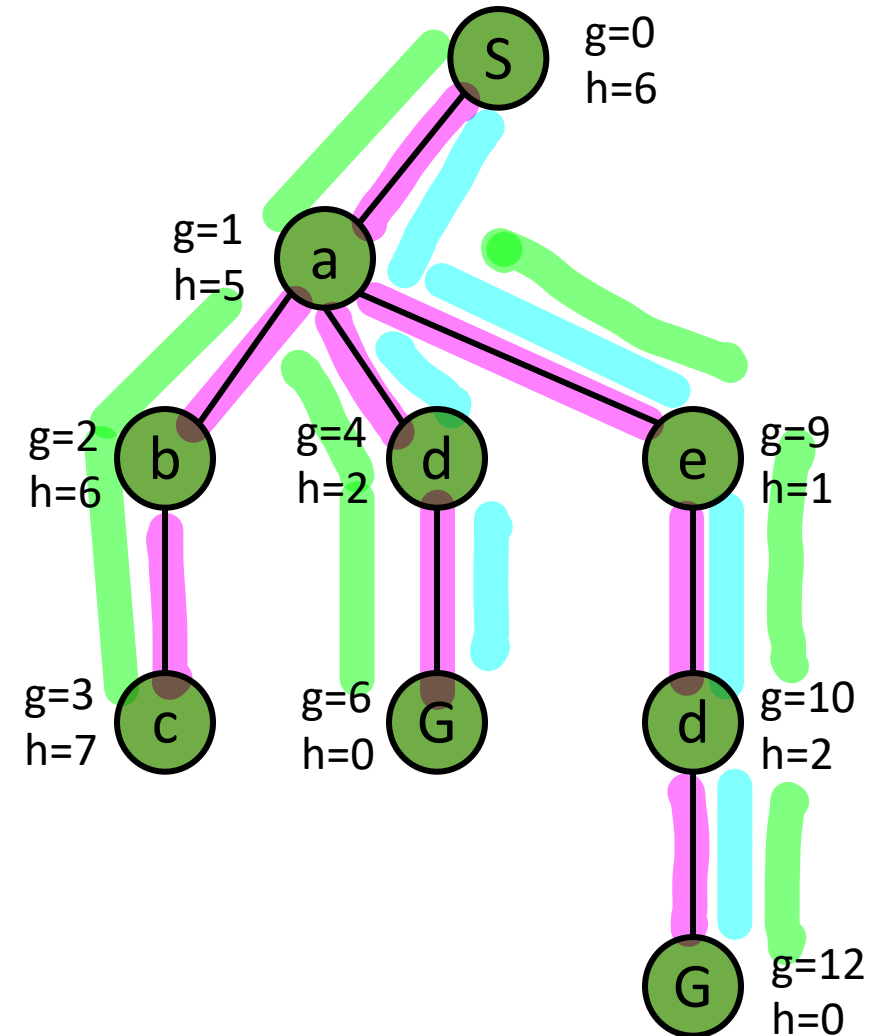
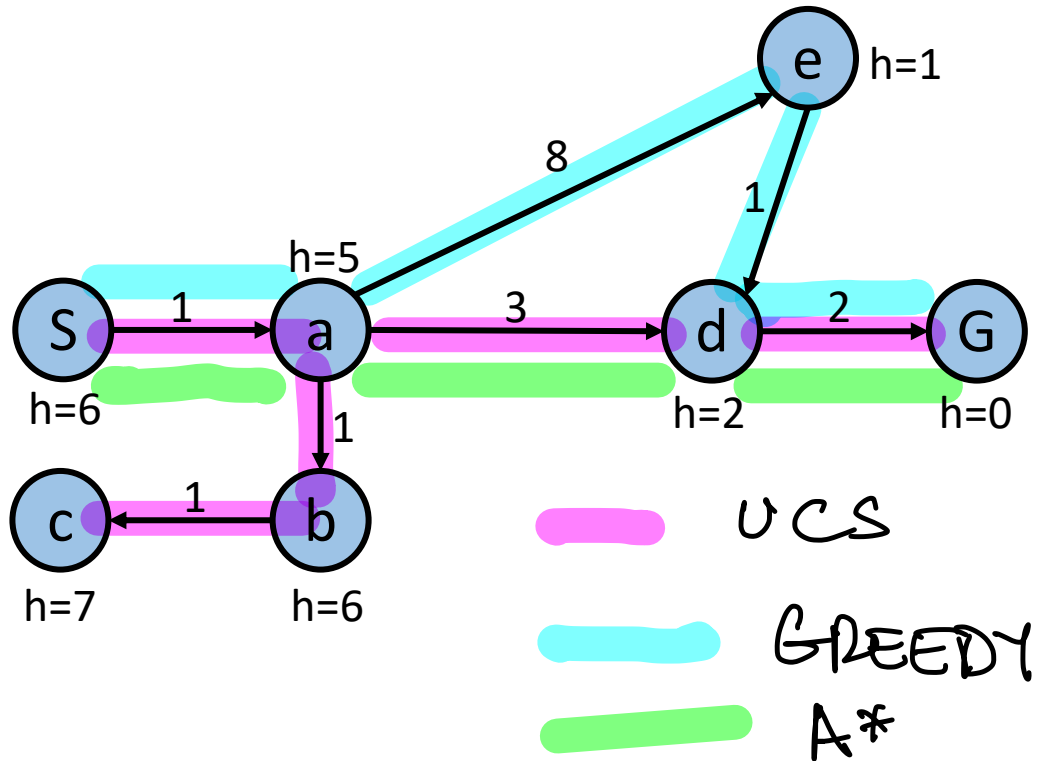
$$f(n) = h(n) \quad (\text{estimated cost to get from } n \text{ to goal})$$

A*:

$$f(n) = g(n) + h(n) \quad (\text{estimated total cost of cheapest solution through } n)$$

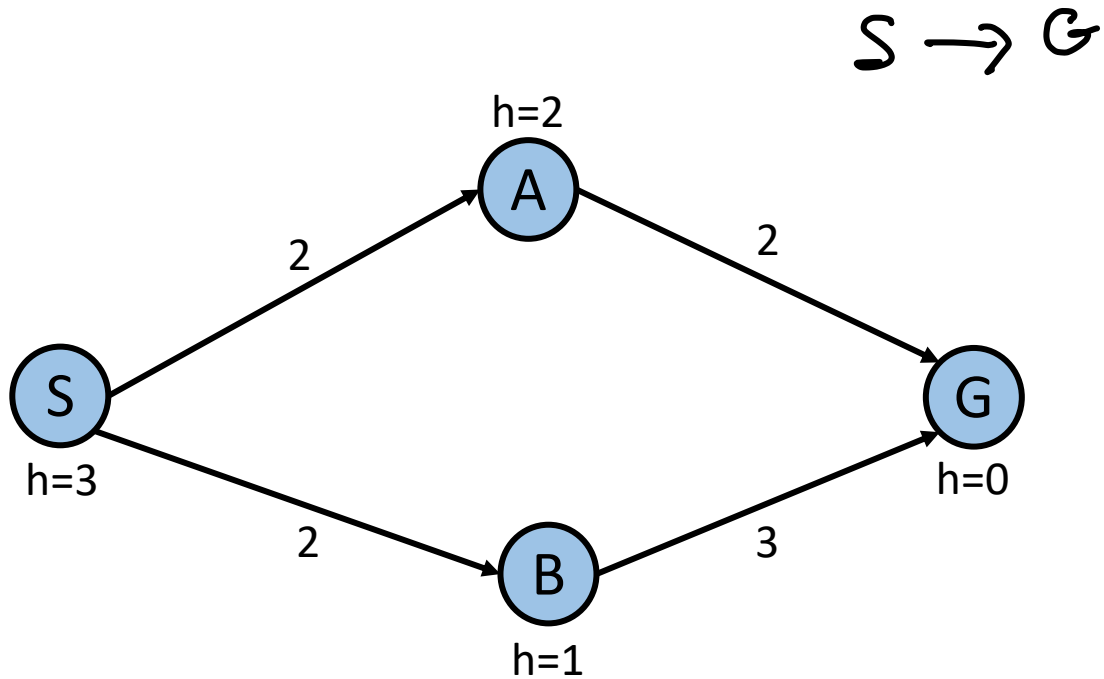
A* Search

Example: Compare Uniform Cost, Greedy Search, and A* on the graph below. S→G



A* Search

Example: When should A* search terminate?



1. $S(\emptyset)$

2. $A(4)$
 $B(3)$

3. $A(4)$
 $G(5)$

4. $G(4)$

5. $G(4)$

$S \rightarrow A, B$

$$A = (2+2) = 4$$

$$B = (2+1) = 3$$

$B \rightarrow G$

$$G = (2+3+0) = 5$$

$A \rightarrow G$

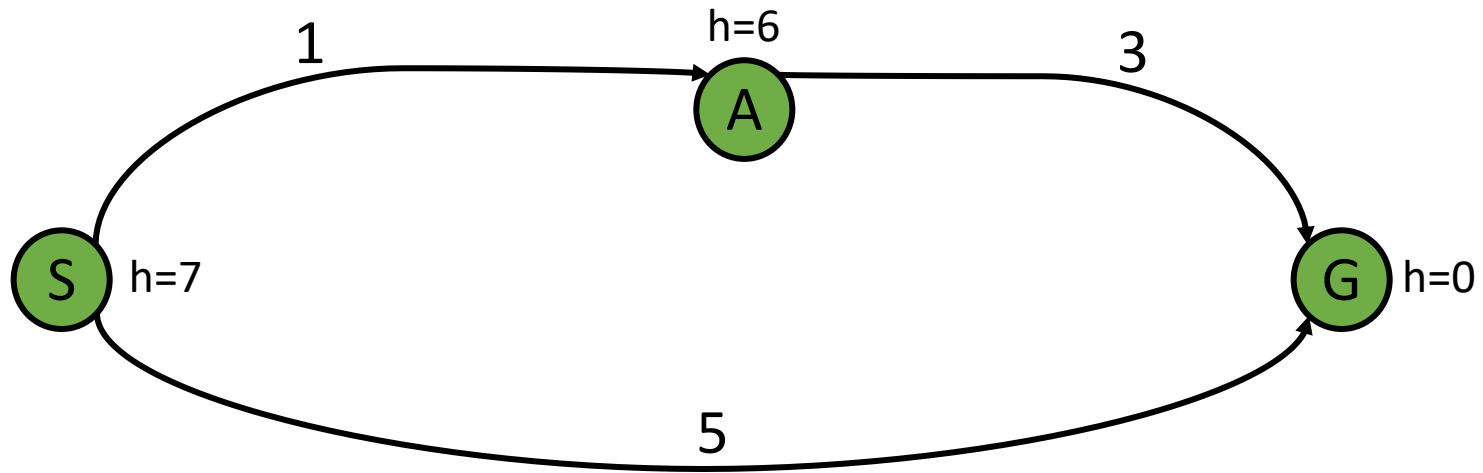
$$G = (2+2+0) =$$

4

Done

A* Search

Is A* optimal?



A* Search

Consistent: for every node n and successor n' of n , generated by some action a , the estimated cost of reaching the goal from n is no greater than the step cost from n to n' , plus the estimated cost of reaching the goal from n'

- That is: $h(n) \leq c(n, a, n') + h(n')$
- General **triangle inequality** between n , n' , and the goal

A heuristic h is **admissible** (optimistic) if $0 \leq h(n) \leq h^*(n)$, where $h^*(n)$ is the true cost to the nearest goal.

Optimality

Conditions for Optimality: Admissibility & Consistency

- $h(n)$ must be **admissible** - an admissible heuristic is one that never overestimates the cost to reach the goal.
- $h(n)$ is **consistent** if, for every node n and every successor n' of n generated by any action a , the estimated cost of reaching the goal from n is no greater than the step cost of getting to n' plus the estimated cost of reaching the goal from n' :

$$h(n) \leq c(n, a, n') + h(n')$$

Optimality of A* Search

A* is **optimally efficient** for any given heuristic: No other optimal algorithm is guaranteed to expand fewer nodes than A*

- Recall: A* expands all nodes with $f(n) < C^*$, where C^* is the cost of the optimal solution path.
- Any algorithm that does not expand all nodes with $f(n) < C^*$ risks missing a better solution path.

Optimality of A* Search

A* (graph) is optimal if the heuristic $h(n)$ is consistent.

Based on two key facts:

1. If $h(n)$ is consistent, then the values of $f(n)$ along any path are nondecreasing.
2. Whenever A* selects a node n for expansion, the optimal path to that node has been found.

Optimality of A* Search

A* (graph) is optimal if the heuristic $h(n)$ is consistent.

Based on two key facts:

1. If $h(n)$ is consistent, then the values of $f(n)$ along any path are nondecreasing.
2. Whenever A* selects a node n for expansion, the optimal path to that node has been found.

➤ So the first goal node to be expanded took the lowest-cost path, and all later goal node expansions are at least as expensive.

Optimality of A* Search

So A* is **optimal**, **complete**, and **optimally efficient**.

Why do we even care about other search algorithms?

- **Number of nodes** to expand along the goal contour is still **exponential** in depth of solution/length of solution path.
- Absolute error: $\Delta := h^* - h$
 - h^* = actual cost from root to goal
 - h = heuristic you used
- Relative error: $\epsilon := (h^* - h)/h^*$

A* Search

Search only works when:

- domain is fully observable
- domain must be known
- domain must be deterministic
- domain must be static

A* Search

Complexity depends strongly on state space characterization

- Single goal, tree, reversible actions $\rightarrow O(b^\Delta)$, or $O(b^{\epsilon d})$ with constant step costs (d is solution depth)

Δ typically is proportional to the path cost h^* , so ϵ is pretty much constant (or growing with d), and we can rewrite: $O((b^\epsilon)^d)$

→ The effective branching factor is really b^ϵ .

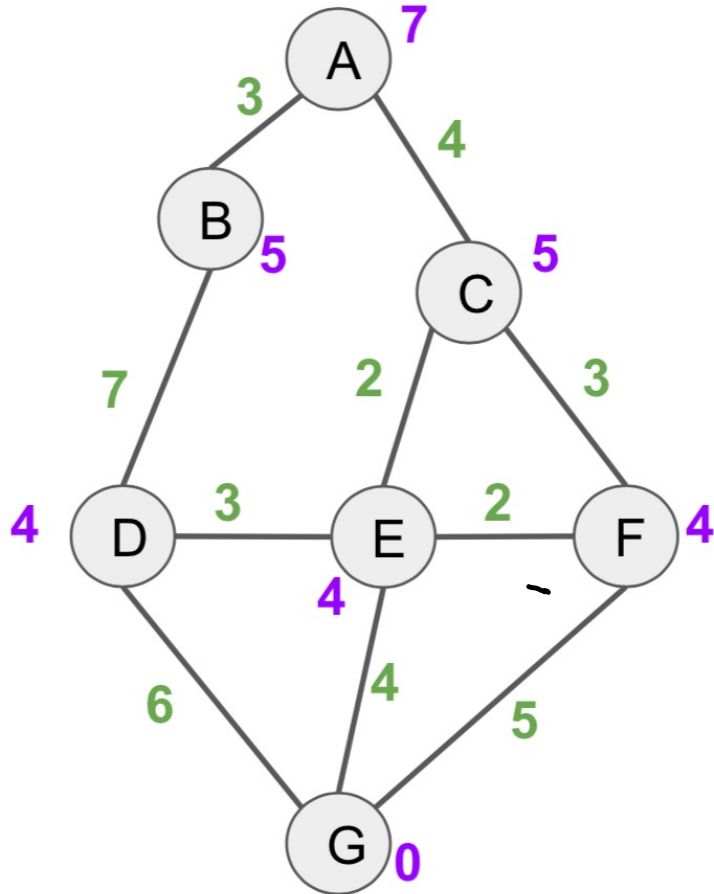
→ Important to choose as good of a heuristic as we can.

- Many goal states/near-goal states can be a problem -- need to expand a **lot** of branches.

A* Search

A* Search:

- Find the minimum cost path from A to G
- $h(n)$ values are given in **purple**
- Step costs are given in **green**



A → G

$$f(n) = g(n) + h(n)$$

path cost + heuristic

	<u>F</u>	<u>E</u>	<u>ACTION</u>
1.	A(7)		
2.	B(8), C(9)	A	A → B(3+5), C(4+5)
3.	C(9) D(14)	A, B	B → D(3+7+4)
4.	D(14), E(10), F(11)	A, B, C	C → E(4+2+4), F(4+3+4)
5.	D(14), F(11), G(10)	A, B, C, E	E → F(4+2+2+4), G(4+2+4+0)
6.	D(14), F(11)	A, B, C, E, G	G → F(4+2+4+5 +4) = 19 X DONE!

A* Search

Example: Use A* search to find a route from Chicago to Providence.

Distances:

Chicago->833

Detroit->597

Indianapolis->783

Columbus->618

Cleveland->530

Pittsburgh->456

Buffalo->388

Syracuse->256

Baltimore->324

Philadelphia->235

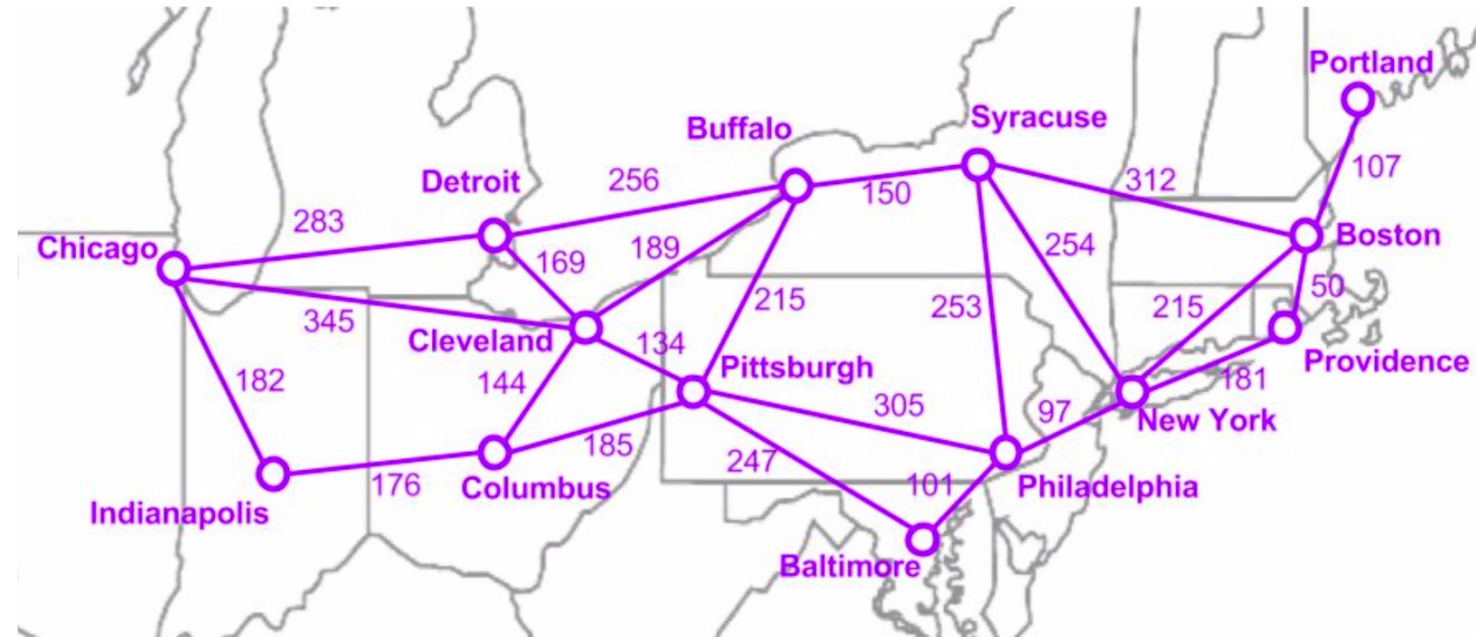
New York->155

Boston->41

Portland->140

$h(n)$ = straight-line distance to Providence

$g(n)$ = Path cost so far



$$h(\text{Chicago}) = \underline{833}$$

<u>F</u>	<u>E</u>
1. Chi (833)	0
2. Det (880), Cle (875), Ind (965)	Chi
3. Buf (922), Pit (935), Col (1107), Det (880), Ind (965)	Chi Cle
4. Buf (922), Pit (935), Col (1107), Ind (965)	Chi Cle Det

ACTION

Chi \rightarrow Det, Cle, Ind

$$\begin{aligned} \text{Det} &= 283 + 597 = 880 \\ \text{Cle} &= 345 + 530 = 875 \\ \text{Ind} &= 182 + 783 = 965 \end{aligned}$$

Cle \rightarrow Buf, Pit, Col

$$\begin{aligned} \text{Buf} &= 345 + 189 + 388 = 922 \\ \text{Pit} &= 345 + 134 + 456 = 935 \\ \text{Col} &= 345 + 144 + 618 = 1107 \end{aligned}$$

Det \rightarrow ~~Chi~~, ~~Cle~~, Buf

$$\text{Buf} = 345 + 169 + 256 + 388 = 1158 \times$$

5. Pit (935) Chi
 Ind (965) Cle
 Col (1107) Det
 Syn (940) Buf

Buf \rightarrow ~~Det~~, Pit, Syn.

$$\text{Pit} = 345 + 189 + 215 + 486 = 1205 \quad X$$

$$\text{Syn} = 345 + 189 + 150 + 256 = 940$$

6. Syn (940) Chi
 Col (1107) Cle
 Ind (965) Det
 Phi (1109) Buf
 Bal (1050) Pit

Pit \rightarrow Phi, Bal, ~~Buf~~

$$\text{Phi} = 345 + 134 + 305 + 235 = 1019$$

$$\text{Bal} = 345 + 134 + 247 + 324 = 1050$$

7. Ind (965) Chi
 Col (1107) Cle
 Phi (1109) Det
 Bal (1050) Buf
 Bor (1042) Pit
 New (1048) Syn
 Phi (1177)

Syn \rightarrow Bor, New, Phi

$$\text{Bor} = 283 + 256 + 150 + 312 + 41 = 1042$$

$$\text{New} = 283 + 256 + 150 + 254 + 155 = 1098$$

$$\text{Phi} = 283 + 256 + 150 + 253 + 235 = 1177 \quad X$$

8.

Col (976)

Phi (1109)

Bal (1050)

Boz (1042)

New (1098)

Chi

Cle

Act

Buf

Pit

Syr

Ind

Ind \rightarrow Col, ~~Chi~~

$$\text{Col} = 182 + 176 + 618 = 976$$

9.

Phi (1109)

Bal (1050)

Boz (1042)

New (1098)

Chi, Cle,

Pit, Det,

Buf, Syr,

Ind, Col

Col \rightarrow ~~Cle~~, ~~Pit~~

10.

Phi (1109)

Bal (1050)

New (1098)

Pro (1051)

Port (1248)

Boz \rightarrow Pro, Port, ~~Syr~~

$$\text{Pro} = 283 + 256 + 150 + 312 + 80 = 1081$$

$$\text{Port} = 283 + 256 + 150 + 312 + 107 + 140 = 1248$$

11.

Phi ¹⁰⁶² (1109)

New (1098)

Pro (1051)

Port (1248)

Bal \rightarrow Phi, ~~Det~~

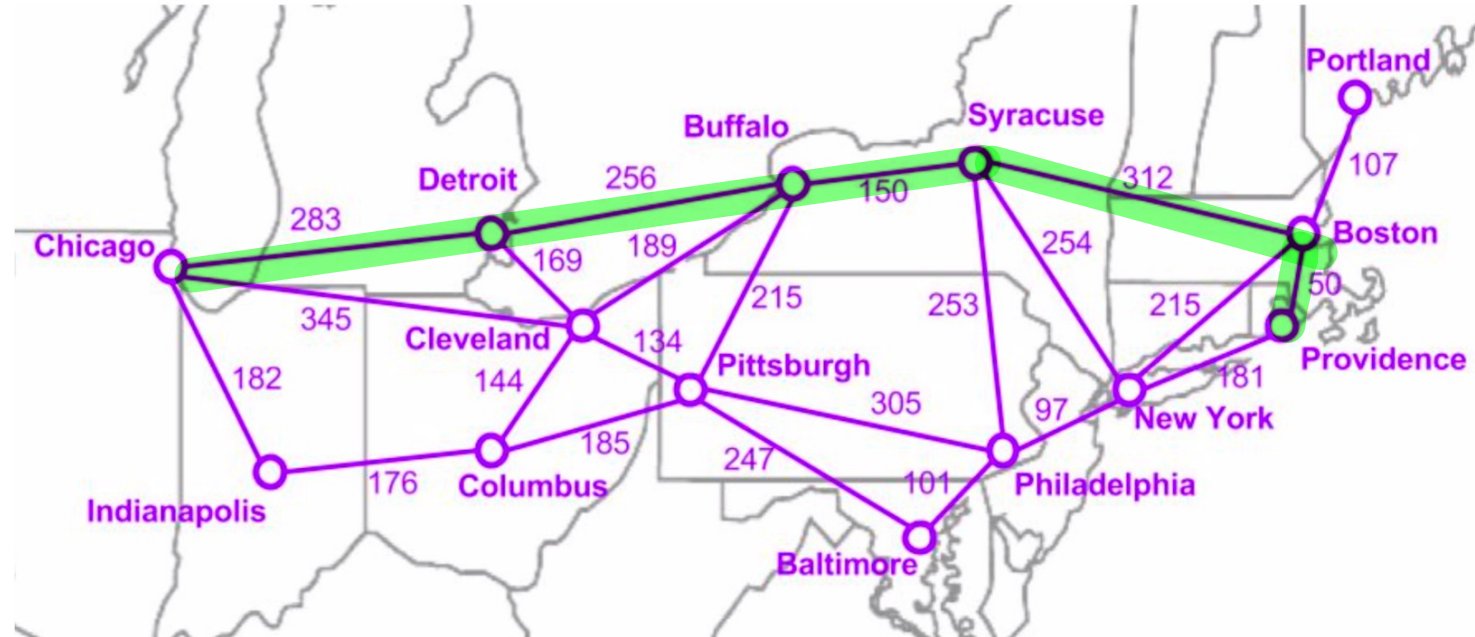
$$Phi = 345 + 134 + 247 + 101 + 235 = 1062$$

12.

Pro \rightarrow Done

Path: Chi \rightarrow Det \rightarrow Buf \rightarrow Sep \rightarrow Bot \rightarrow Providence

Portland->140

$$g(n) = \text{Path cost so far}$$


PATH DISTANCE ≈ 1051 miles