

# CSE 373: More on Dijkstra's algorithm

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Wednesday, Feb 21, 2018

# Dijkstra's algorithm

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2. Set our starting node's cost to 0

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**Idea:** *Greedily* pick node with smallest cost, then update everything possible. Repeat.

# Dijkstra's algorithm

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# Dijkstra's algorithm

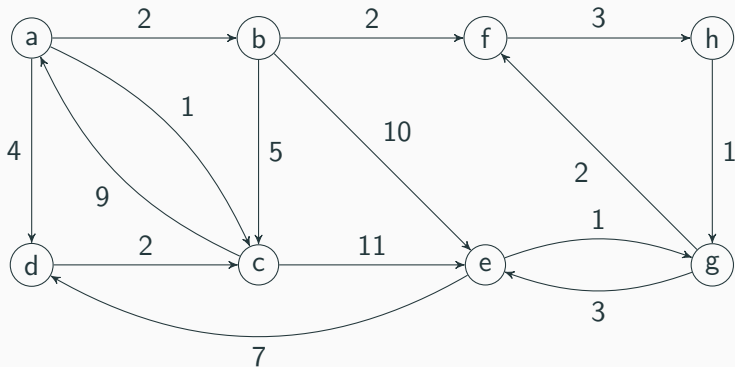
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**Pronunciation:** DYKE-struh (“dijk” rhymes with “bike”)

# Dijkstra's algorithm

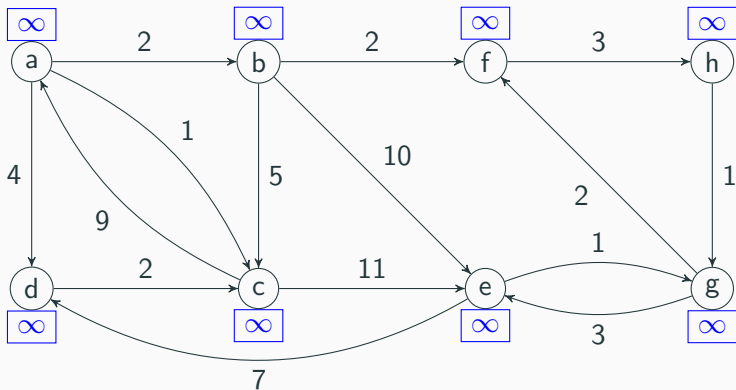
Suppose we start at vertex "a":





# Dijkstra's algorithm

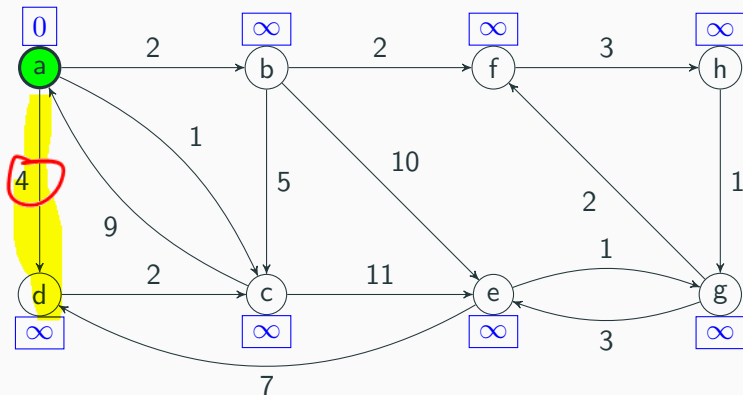
Suppose we start at vertex "a":



We initially assign all nodes a cost of infinity.

# Dijkstra's algorithm

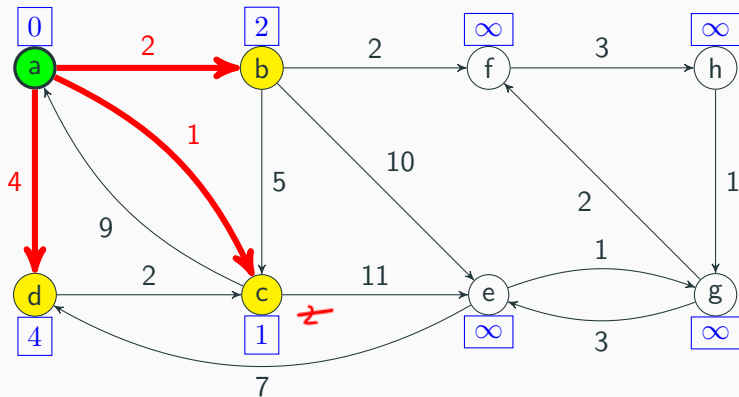
Suppose we start at vertex "a":



Next, assign the starting node a cost of 0.

# Dijkstra's algorithm

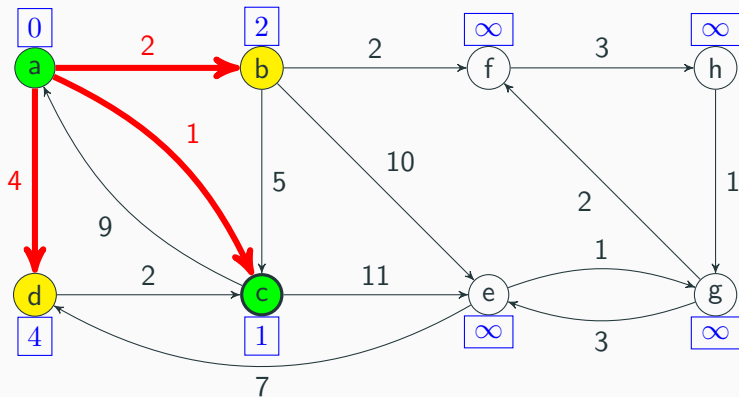
Suppose we start at vertex "a":



Next, update all adjacent node costs as well as the backpointers.

# Dijkstra's algorithm

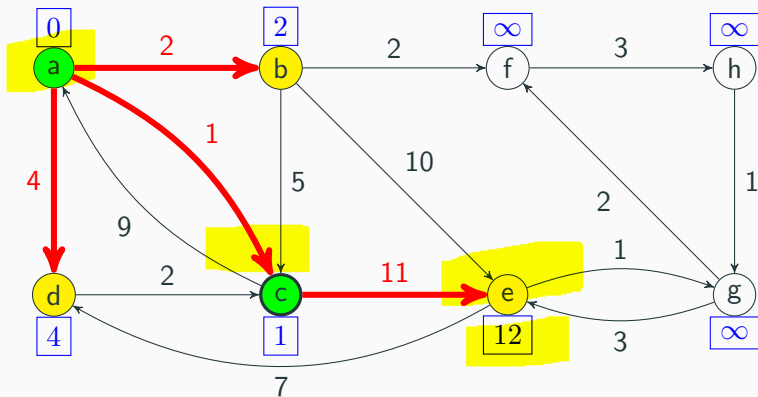
Suppose we start at vertex "a":



The pending node with the smallest cost is c, so we visit that next.

# Dijkstra's algorithm

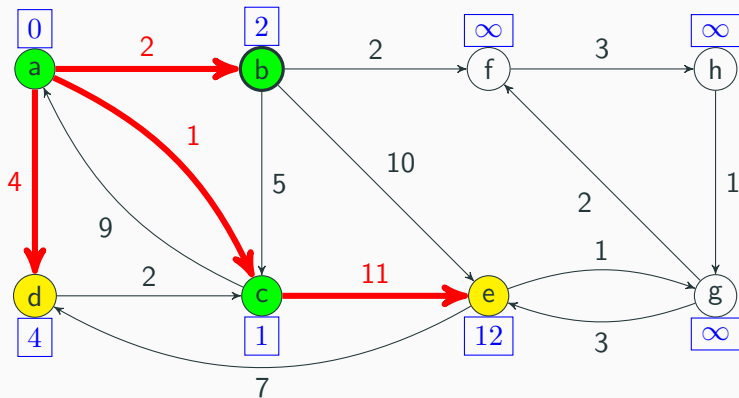
Suppose we start at vertex "a":



We consider all adjacent nodes.  $a$  is fixed, so we only need to update  $e$ . Note the new cost of  $e$  is the sum of the weights for  $a - c$  and  $c - e$ .

# Dijkstra's algorithm

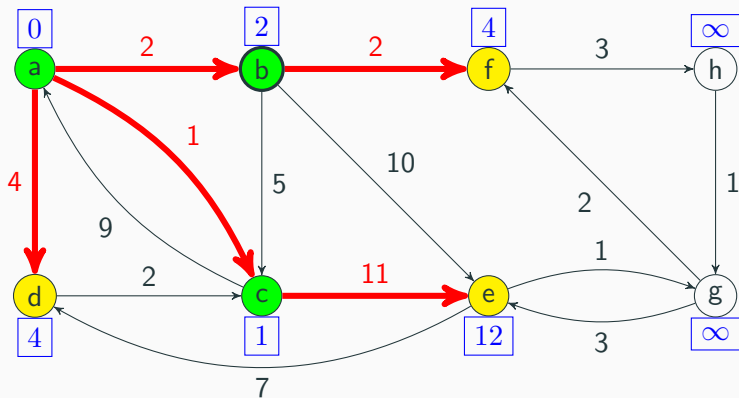
Suppose we start at vertex "a":



*b* is the next pending node with smallest cost.

# Dijkstra's algorithm

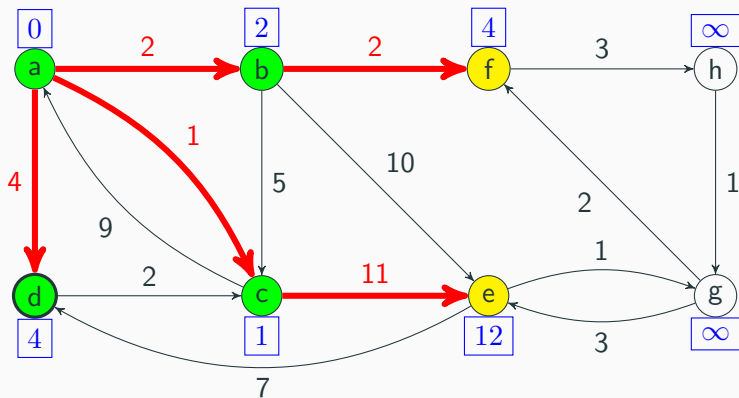
Suppose we start at vertex "a":



The adjacent nodes are c, e, and f. The only node where we can update the cost is f. Note the route a – b – e has the same cost as a – c – e, so there's no point in updating the backpointer to e.

# Dijkstra's algorithm

Suppose we start at vertex "a":

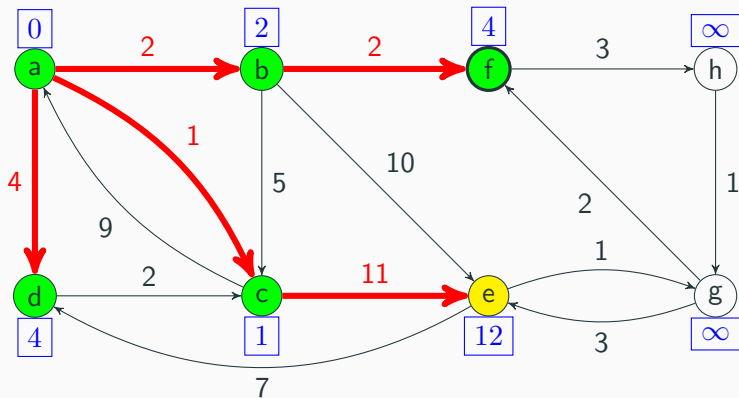


Both  $d$  and  $f$  have the same cost, so let's (arbitrarily) pick  $d$  next. Note that we can't adjust any of our neighbors.



# Dijkstra's algorithm

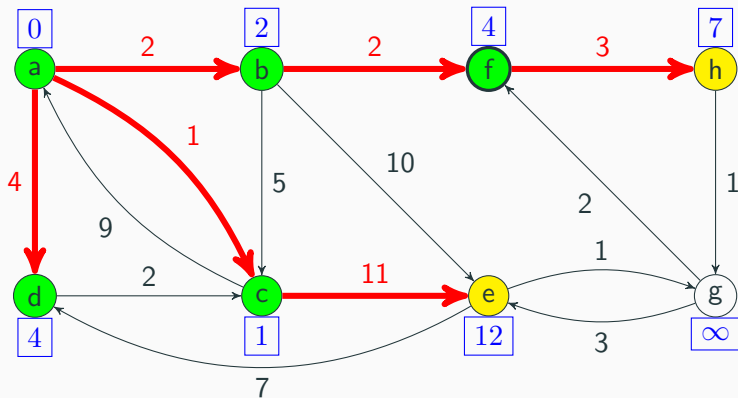
Suppose we start at vertex "a":



Next up is *f*.

# Dijkstra's algorithm

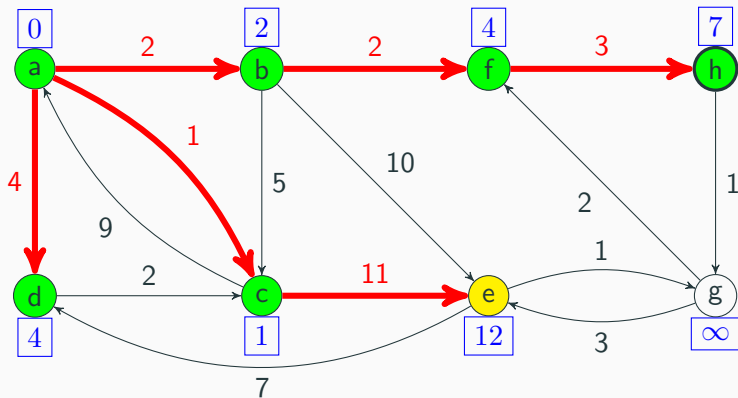
Suppose we start at vertex "a":



The only neighbor we is  $h$ .

# Dijkstra's algorithm

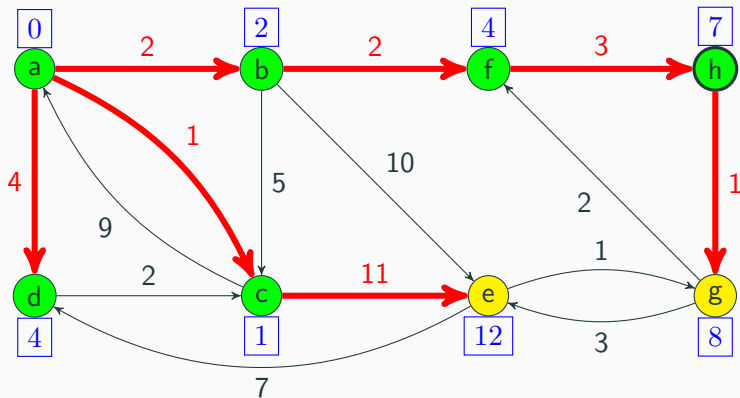
Suppose we start at vertex "a":



*h* has the smallest cost now.

# Dijkstra's algorithm

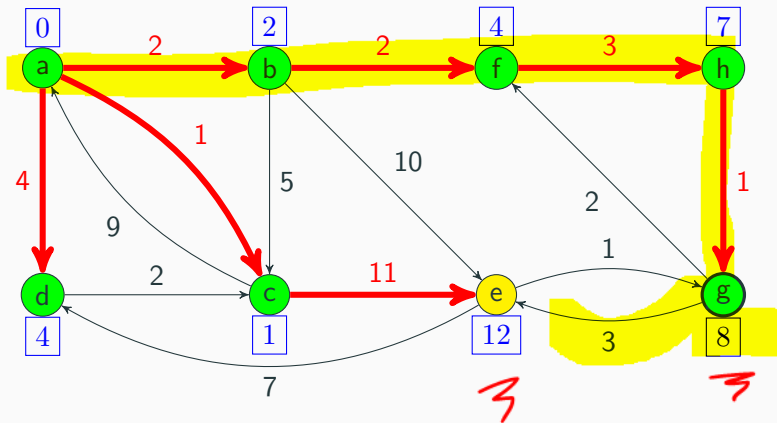
Suppose we start at vertex "a":



We update g.

# Dijkstra's algorithm

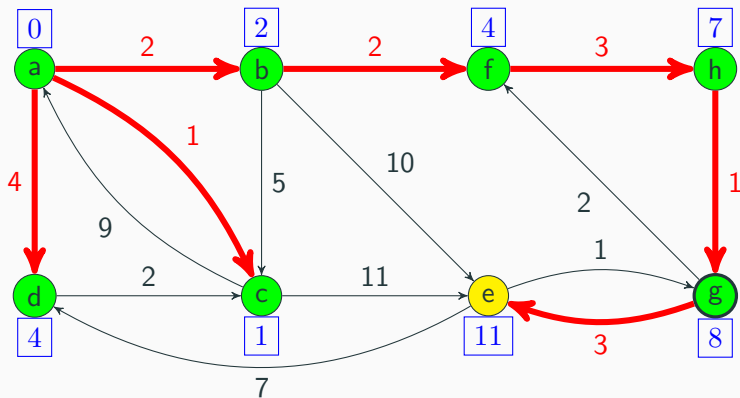
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Next up is g.

# Dijkstra's algorithm

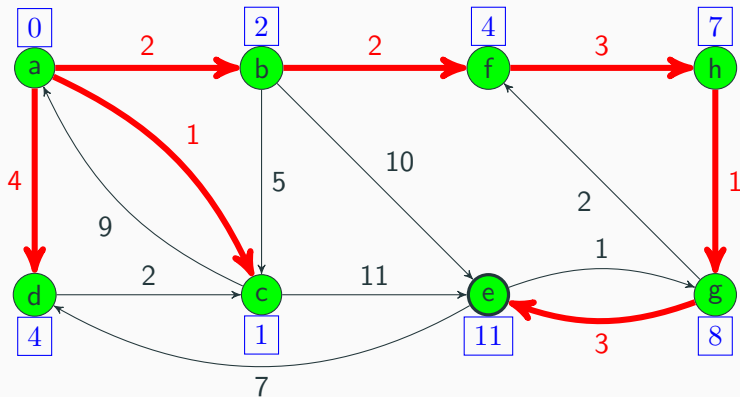
Suppose we start at vertex "a":



The two adjacent nodes are *f* and *e*. *f* is fixed so we leave it alone. We however will update *e*: our current route is cheaper than the previous route, so we update both the cost and the backpointer

# Dijkstra's algorithm

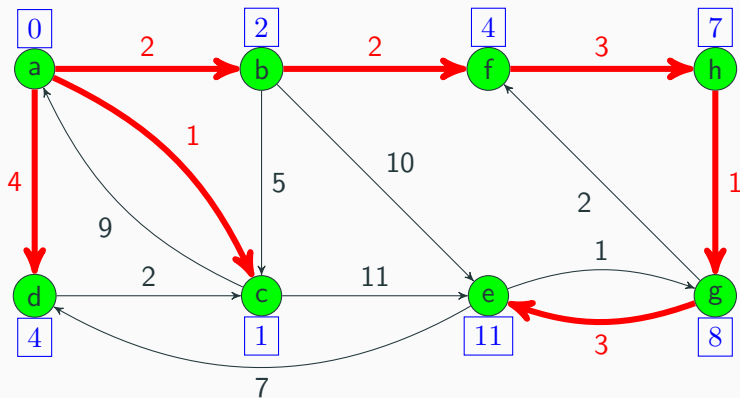
Suppose we start at vertex "a":



The last pending node is e. We visit it, and check for any unfixed adjacent nodes (there are none).

# Dijkstra's algorithm

Suppose we start at vertex "a":



And we're done! Now, to find the shortest path, from a to a node, start at the end, trace the red arrows backwards, and reverse the list.



## Core idea in simplified pseudocode:

```
def dijkstra(start):  
    for (v : vertices):  
        set cost(v) to infinity  
    set cost(start) to 0  
  
    while (we still have unvisited nodes):  
        current = get next smallest node  
  
        for (edge : current.getOutEdges()):  
            newCost = min(cost(current) + edge.cost, cost(edge.dest))  
            update cost(edge.dest) to newCost, update backpointers, etc  
  
    return backpointers dictionary
```

# Dijkstra's algorithm

## One implementation: inserting extra values into heap

```
def dijkstra(start):
    backpointers = empty Dictionary of vertex to vertex
    costs = Dictionary of vertex to double, initialized to infinity
    visited = empty Set

    heap = new Heap<Node with cost>();
    heap.put([start, 0])
    cost.put(start, 0)
    while (heap is not empty):
        current, currentCost = heap.removeMin()
        skip if visited.contains(current), else visited.add(current)

        for (edge : current.getOutEdges()):
            skip if visited.contains(edge.dest), else visited.add(edge.dest)

            if (newCost < cost.get(edge.dest)):
                cost.put(edge.dest, newCost)
                heap.insert([edge.dest, newCost])
                backpointers.put(edge.dest, current)

    return backpointers dictionary
```

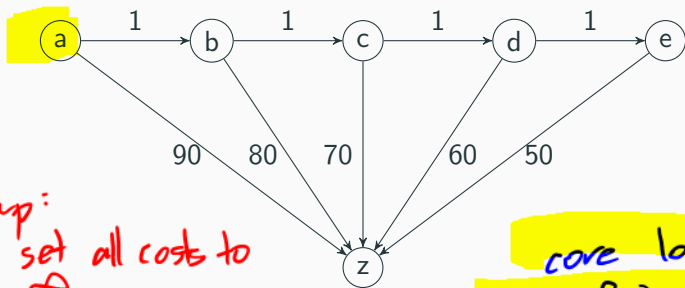
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## Another impl: after implementing decreasePriority

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    backpointers = empty Dictionary of vertex to vertex  
    costs = empty Dictionary of vertex to double  
  
    heap = new Heap<Node with cost>();  
    for (v : vertices):  
        heap.put([v, infinity])  
        costs.put(v, infinity)  
  
    heap.decreasePriority([start, 0])  
    costs.put(start, 0)  
  
    while (heap is not empty):  
        current, currentCost = heap.removeMin()  
  
        for (edge : current.getOutEdges()):  
            newCost = currentCost + edge.cost  
            if (newCost < cost.get(edge.dest)):  
                cost.put(edge.dest, newCost)  
                heap.decreaseKey([edge.dest, newCost])  
                backpointers.put(edge.dest, current)  
  
    return backpointers dictionary
```

## Example

What does Dijkstra's algorithm do when run on vertex *a*?



set up:

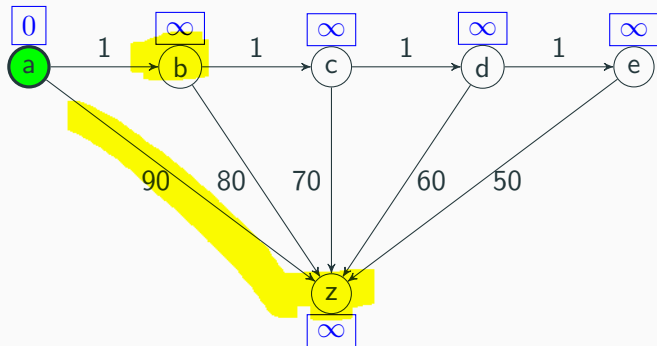
- set all costs to  $\infty$
- set *a*'s cost to 0

core loop

1. find node w/ smallest cost
2. update neighbors
3. repeat

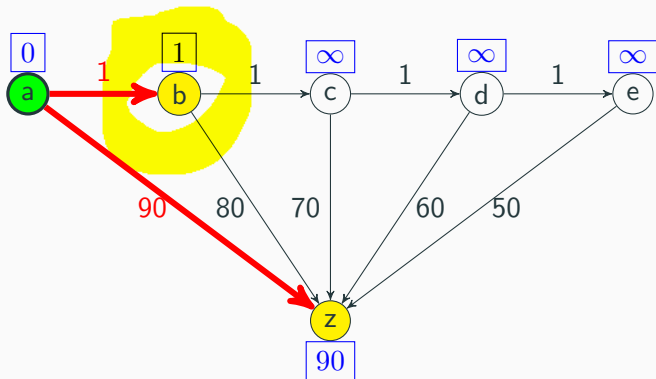
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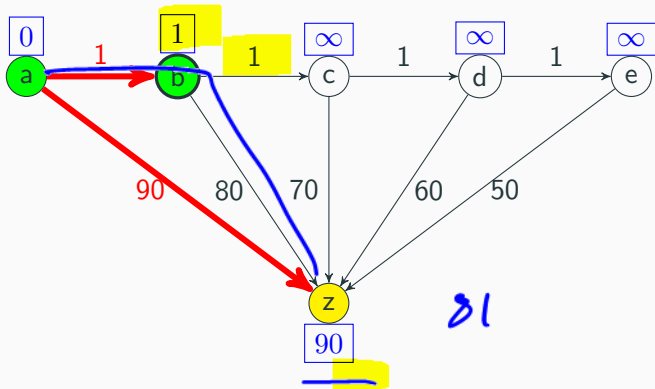
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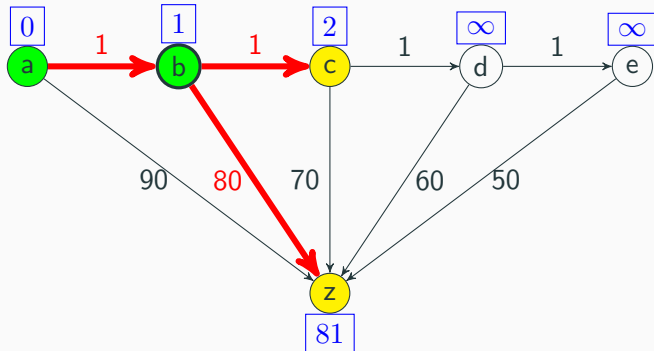
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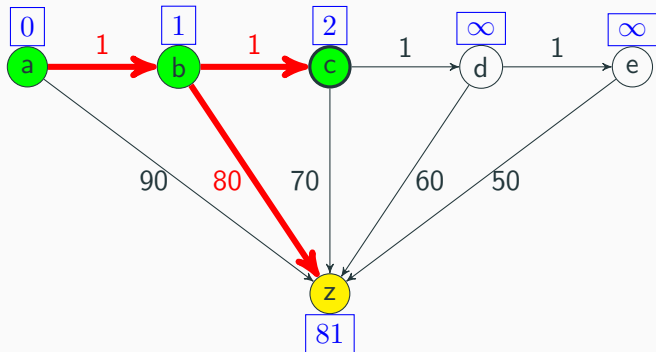
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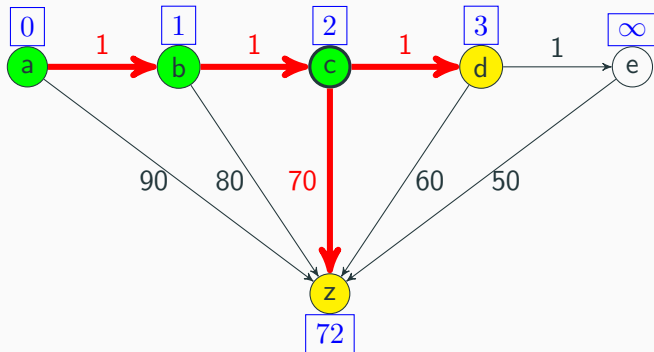
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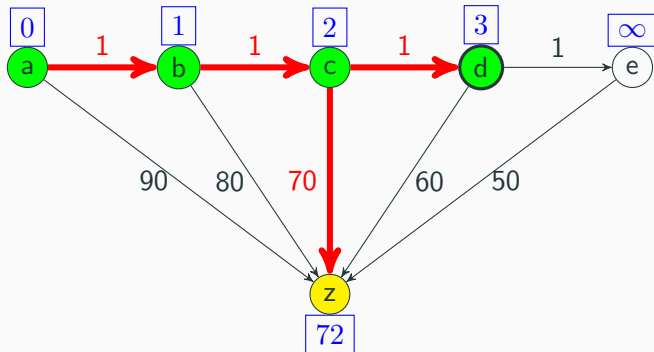
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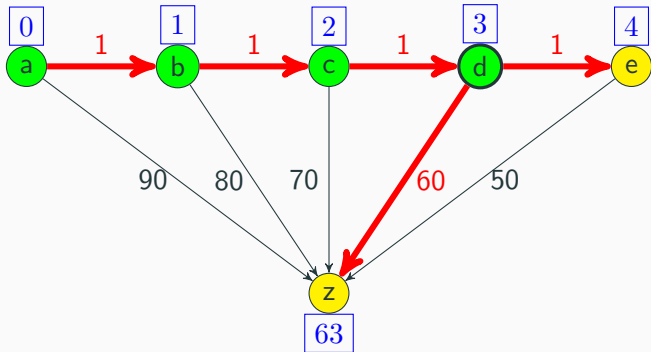
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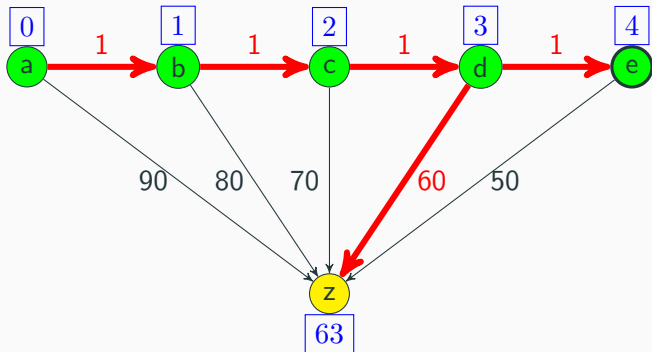
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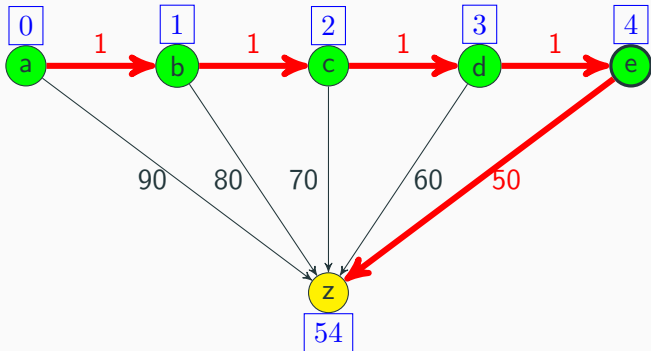
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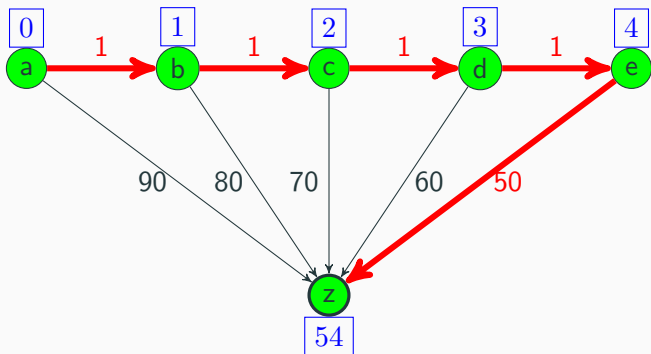
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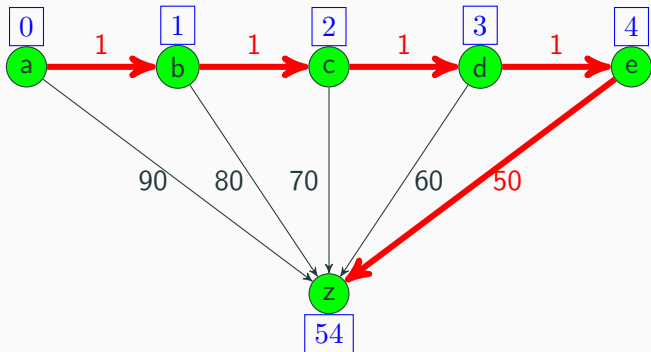
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  - ▶ Project 3, part 1 grades also released later tonight
- Reminder: if you fix the errors in your Friday submission, you can get up to half credit back.

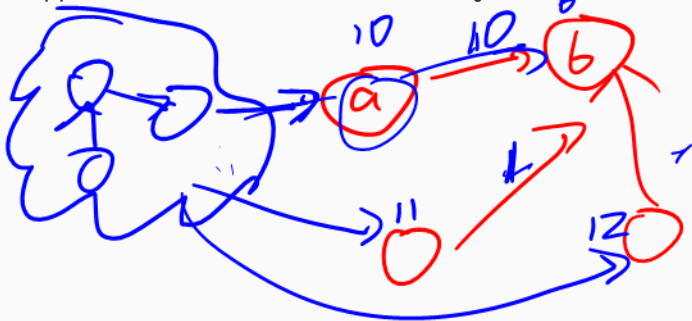
## Misc announcements

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- ▶ If you've emailed me, and you haven't heard back, email me again

## Dijkstra's: why does it work?

Rough intuition:

- Suppose  $a$  is the next unvisited node with the smallest cost. Suppose  $b$  is some unvisited vertex adjacent to  $a$ .



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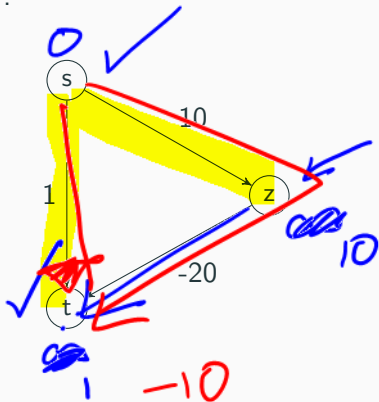
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- ▶ So, picking the shortest node will always accurately update the adjacent nodes.

(Full proof beyond scope of class)

## Dijkstra's: negative edges

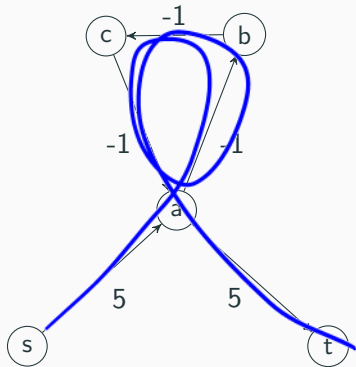
What if we have negative edges?

**Question:** What's the shortest path from  $s$  to  $t$  according to Dijkstra's? In reality?



## Dijkstra's: negative edges

What's the shortest path now?





### Punchline:

- ▶ If there are negative edges, Dijkstra's doesn't work  
(There exist other algorithms that can handle negative edges  
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(Where do negative edges show up? Examples: modeling credit and debit, modeling flow of energy, etc.)

**Question:** what is the worst-case runtime of Dijkstra's algorithm?

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**Strategy 1:** Analyze the code, like we've been doing all quarter

**Strategy 2:** Analyze the algorithm more holistically, like we did for DFS and BFS

## Dijkstra's algorithm: analyzing runtime via code

Consider this (simplified) pseudocode. How do we analyze?


```
def dijkstra(start):  
    for (v : vertices):  
        set cost(v) to infinity  
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    while (we still have unvisited nodes):  
        current = get next smallest node  
  
        for (edge : current.getOutEdges()):  
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Handwritten annotations in blue and red:

- A large blue bracket on the right side of the code, spanning from the `while` loop to the `return` statement, with  $O(|V|)$  written next to it.
- A smaller blue bracket next to the `current = get next smallest node` line, with  $t_s$  written next to it.
- A red arrow points to the `for` loop.
- A blue bracket next to the `for` loop, with  $|E|$  written next to it.
- A blue bracket next to the `update cost` line, with  $t_u$  written next to it.

(Note: let  $t_s$  be the time needed to get the next smallest node, and let  $t_u$  be the time needed to update vertex costs. We'll treat these as unknowns for now.)

$$|V| + |V| \cdot (t_s + |E| \cdot t_u)$$

# Dijkstra's algorithm: analyzing runtime via code

## Things we know:

- ▶ Initialization takes  $\mathcal{O}(|V|)$  time
- ▶ The while loop repeats  $|V|$  times
- ▶ The inner foreach loop repeats  $|E|$  times (???)?
- ▶ The inner foreach loop does  $\mathcal{O}(t_u)$  work per iteration
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Distribute:

$$\mathcal{O}(|V| + |V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

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Distribute:

$$\mathcal{O}(|V| + |V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

The lone  $|V|$  is dominated by  $|V| \cdot t_s$ :

$$\mathcal{O}(|V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

## Dijkstra's algorithm: analyzing runtime

Our runtime:

$$\mathcal{O}(|V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

## Dijkstra's algorithm: analyzing runtime

Our runtime:

$$O(|V| \cdot t_s + \underbrace{|V| \cdot |E|}_{|E|} \cdot t_u)$$

$$|V| = 4$$
$$|E| = 4$$



Question:

Do we really need to update vertex costs  $|V| \cdot |E|$  times?

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while (we still have unvisited nodes):  
    current = get next smallest node  
  
for (edge : current.getOutEdges()):  
    newCost = min(cost(current) + edge.cost, cost(edge.dest))  
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# Dijkstra's algorithm: analyzing runtime

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Original bound:

$$\mathcal{O}(|V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

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    for (edge : current.getOutEdges()):  
        newCost = min(cost(current) + edge.cost, cost(edge.dest))  
        update cost(edge.dest) to newCost, update backpointers, etc
```


## Observations about the foreach loop:

- ▶ We don't know how many times it runs **per** each iteration
- ▶ ...but we do know num times it runs across **all** iterations!

Original bound:

$$\mathcal{O}(|V| \cdot t_s + |V| \cdot |E| \cdot t_u)$$

We update at most once per edge – so, a tighter bound:

$$\mathcal{O}(|V| \cdot t_s + |E| \cdot t_u)$$


## Dijkstra's algorithm: finding and updating nodes

Our runtime so far:

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**Question:** So, what exactly is  $t_s$  and  $t_u$ ?

## Dijkstra's algorithm: finding and updating nodes

Our runtime so far:

$$\mathcal{O}(|V| \cdot t_s + |E| \cdot t_u)$$

**Question:** So, what exactly is  $t_s$  and  $t_u$ ?

**Answer:** Depends on how we store nodes and costs!

## Dijkstra's algorithm: finding and updating nodes

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# Dijkstra's algorithm: finding and updating nodes

**Observation:** there are two operations we care about: finding the node with the min cost, and given a node, updating its cost

## Ideas:

- ▶ Use a binary heaps: lets us find a node with min cost easily
- ▶ Use a dictionary: lets us update the value corresponding to a node easily

# Dijkstra's algorithm: finding and updating nodes

Exercise: fill out this table



Data structure	<u>Remove min (<math>t_s</math>)</u>	<u>Update cost (<math>t_u</math>)</u>
<u>Hash map</u>	$O( V )$	$O(1)$
<u>Sorted array</u>	$O(1)$	$O( V )$
<u>AVL tree</u>	$O(\log  V )$	$O(\log  V )$
<u>Binary heap</u>	$O(\log  V )$	$O( V )$

## Dijkstra's algorithm: finding and updating nodes

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The AVL version looks actually pretty reasonable

## Dijkstra's algorithm: finding and updating nodes

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- ▶ Implementing **removeMin**:  
Run the standard removeMin heap algorithm. As we swap nodes, add some extra code to keep the hash map up-to-date. This is still  $\mathcal{O}(\log(n))$ .
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Run the standard removeMin heap algorithm. As we swap nodes, add some extra code to keep the hash map up-to-date. This is still  $\mathcal{O}(\log(n))$ .
- ▶ Implementing **updateCost**:  
Use the hash map to get the index of the given node. Run percolateUp, updating the hash map as we go. This is still  $\mathcal{O}(\log(n))$ .

## Dijkstra's algorithm: finding and updating nodes

Data structure	removeMin ( $t_s$ )	updateCost ( $t_u$ )
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Sorted array	$\mathcal{O}(1)$	$\mathcal{O}( V )$
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Binary heap	$\mathcal{O}(\log( V ))$	$\mathcal{O}( V )$
“Hybrid” binary heap		

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## Dijkstra's algorithm: finding and updating nodes

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AVL tree	$\mathcal{O}(\log( V ))$	$\mathcal{O}(\log( V ))$
Binary heap	$\mathcal{O}(\log( V ))$	$\mathcal{O}( V )$
“Hybrid” binary heap	$\mathcal{O}(\log( V ))$	$\mathcal{O}(\log( V ))$
Fibonacci heaps	$\mathcal{O}(\log( V ))$	$\mathcal{O}(1)$

Note: Fibonacci heaps are beyond the scope of this class

## Dijkstra's algorithm: finding and updating nodes

**Observation:** Gosh, this all sounds exhausting

What if we replace the binary heap's call to **updateCost** with **insert** and just allow duplicates?

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What if we replace the binary heap's call to **updateCost** with **insert** and just allow duplicates?

Runtime is now  $\mathcal{O}((|V| + |E|) \log(|V| + |E|))$  – the analysis is left as an exercise to the reader.

So, less efficient, but easiest to implement.