

# **CSE 421**

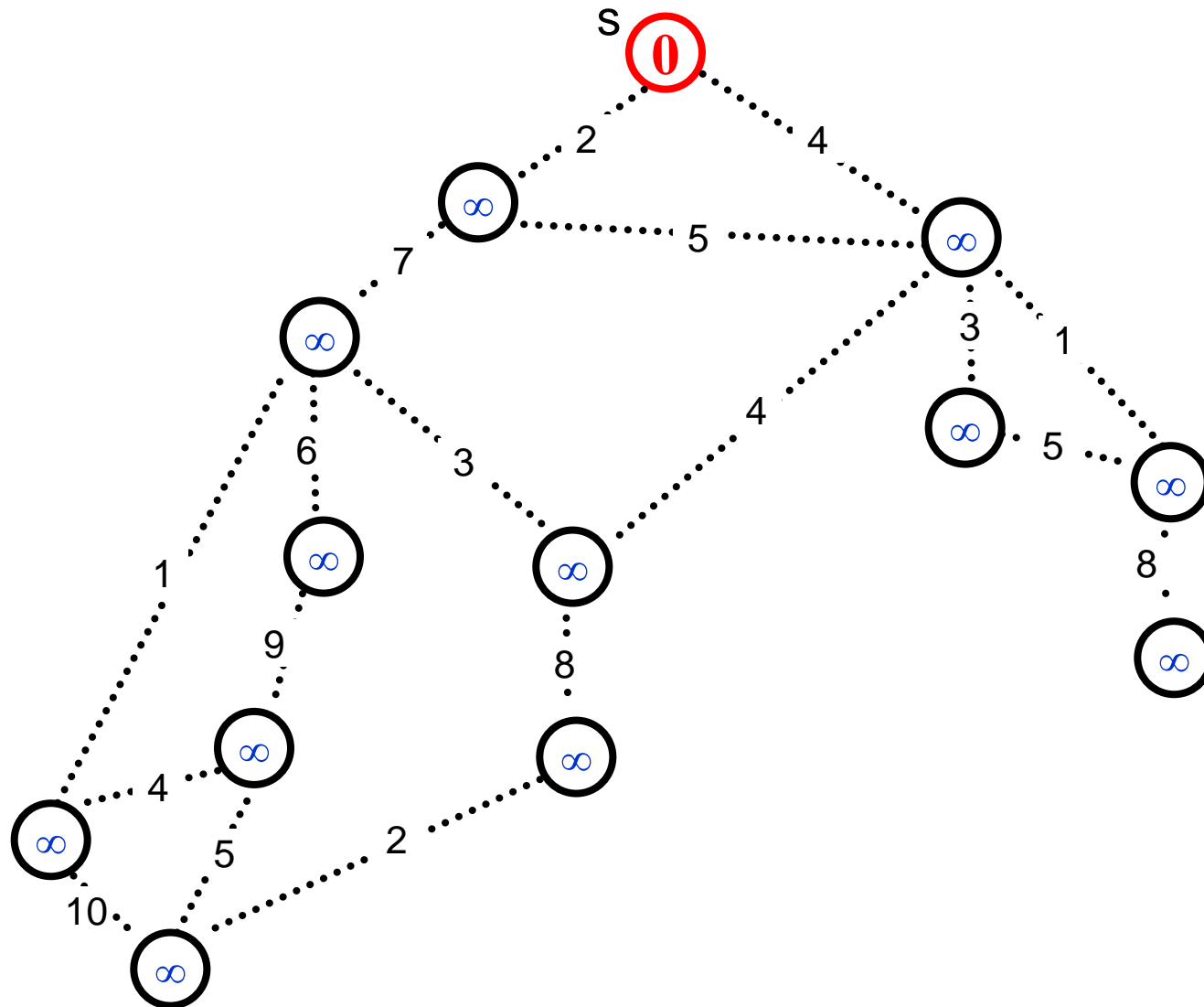
## **Greedy Alg: Union Find/Dijkstra's Alg**

Shayan Oveis Gharan

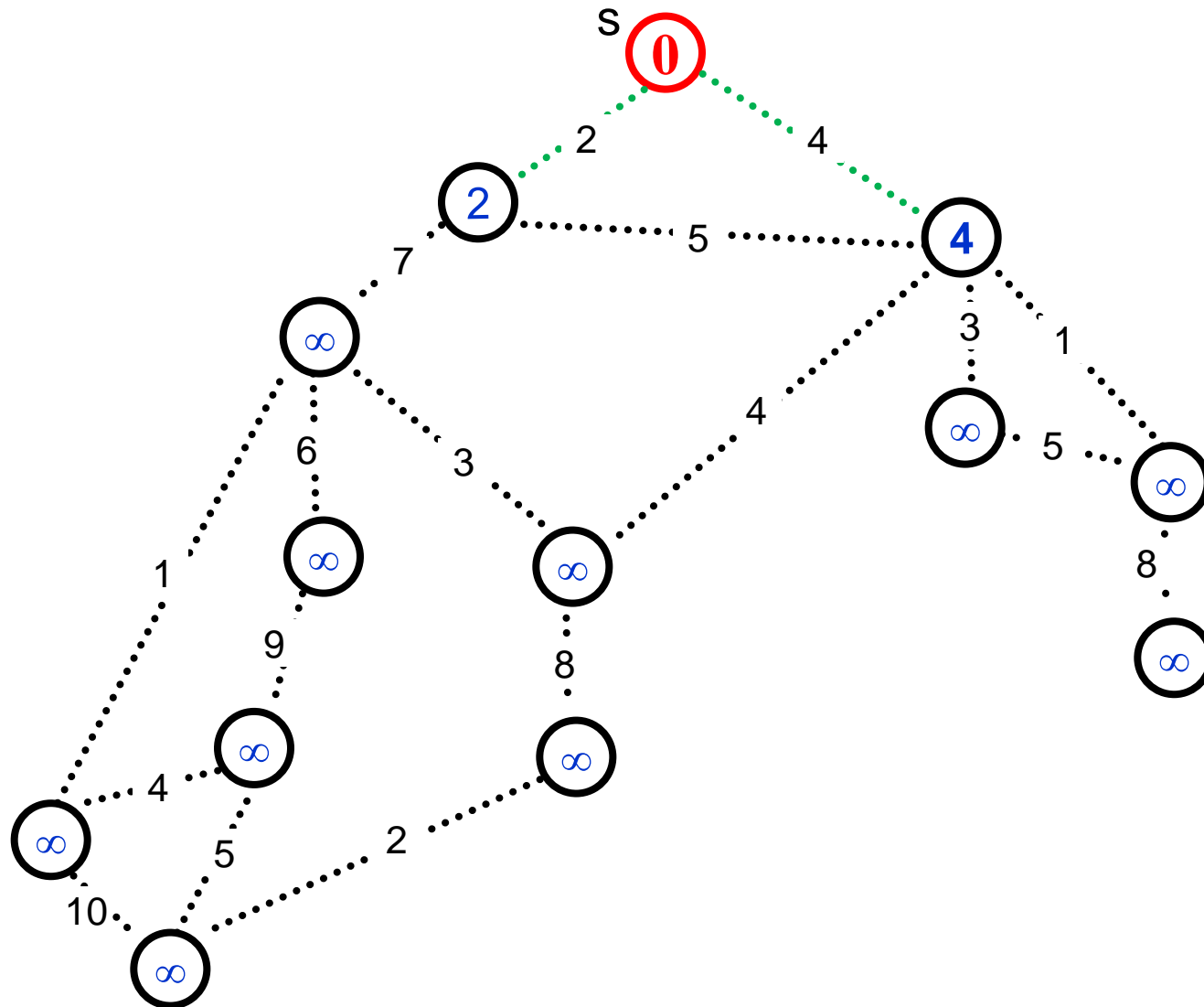
# Dijkstra's Algorithm

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    foreach ( $v \in V$ )  $d[v] \leftarrow \infty$  //This is the key of node v  
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        foreach (edge  $e = (u, v)$  incident to u)  
            if (( $v \notin S$ ) and ( $d[u] + c_e < d[v]$ ))  
                 $d[v] \leftarrow d[u] + c_e$   
                Decrease key of v to  $d[v]$ .  
                Parent(v)  $\leftarrow u$   
    }  
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```

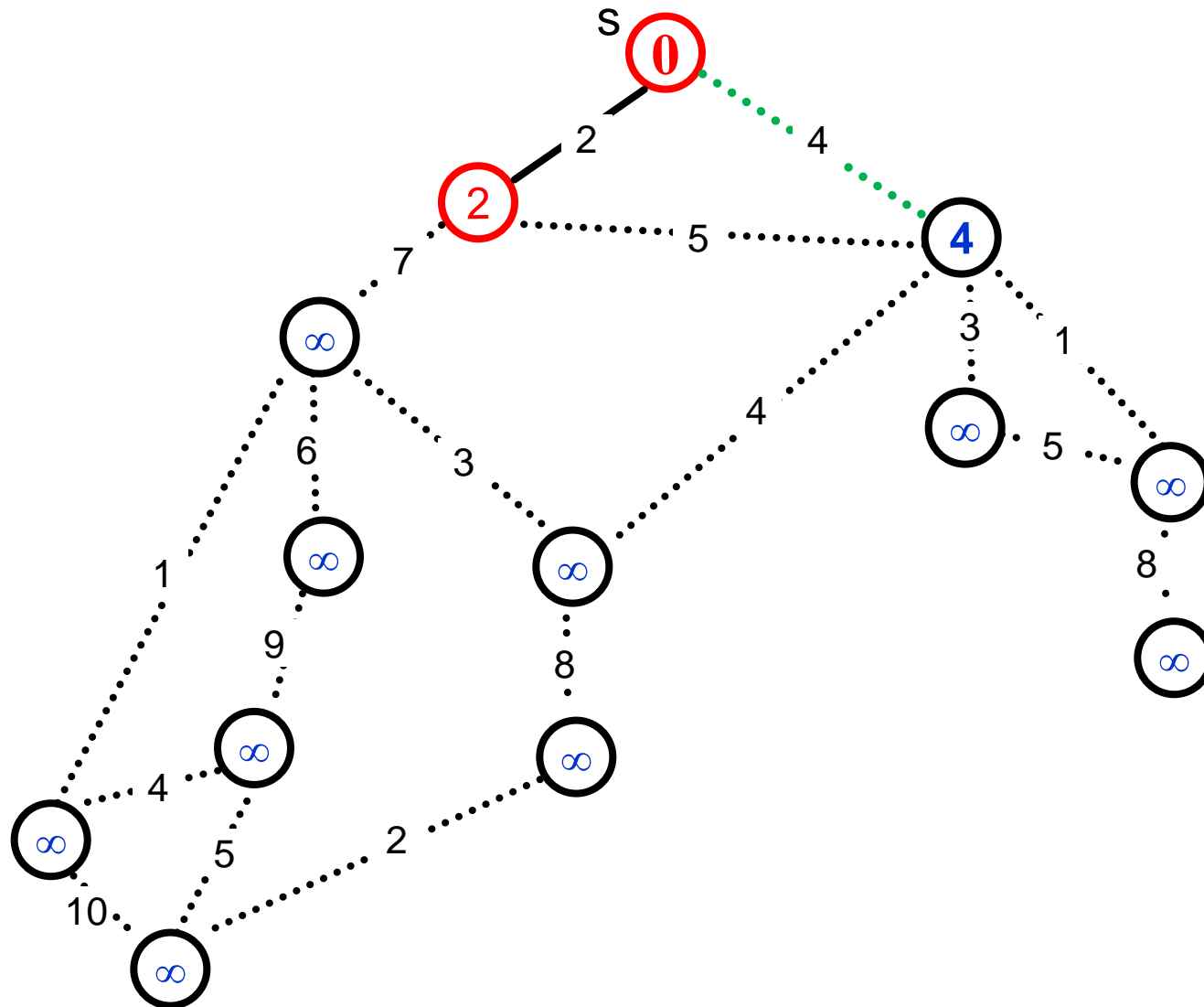
# Dijkstra's Algorithm: Example



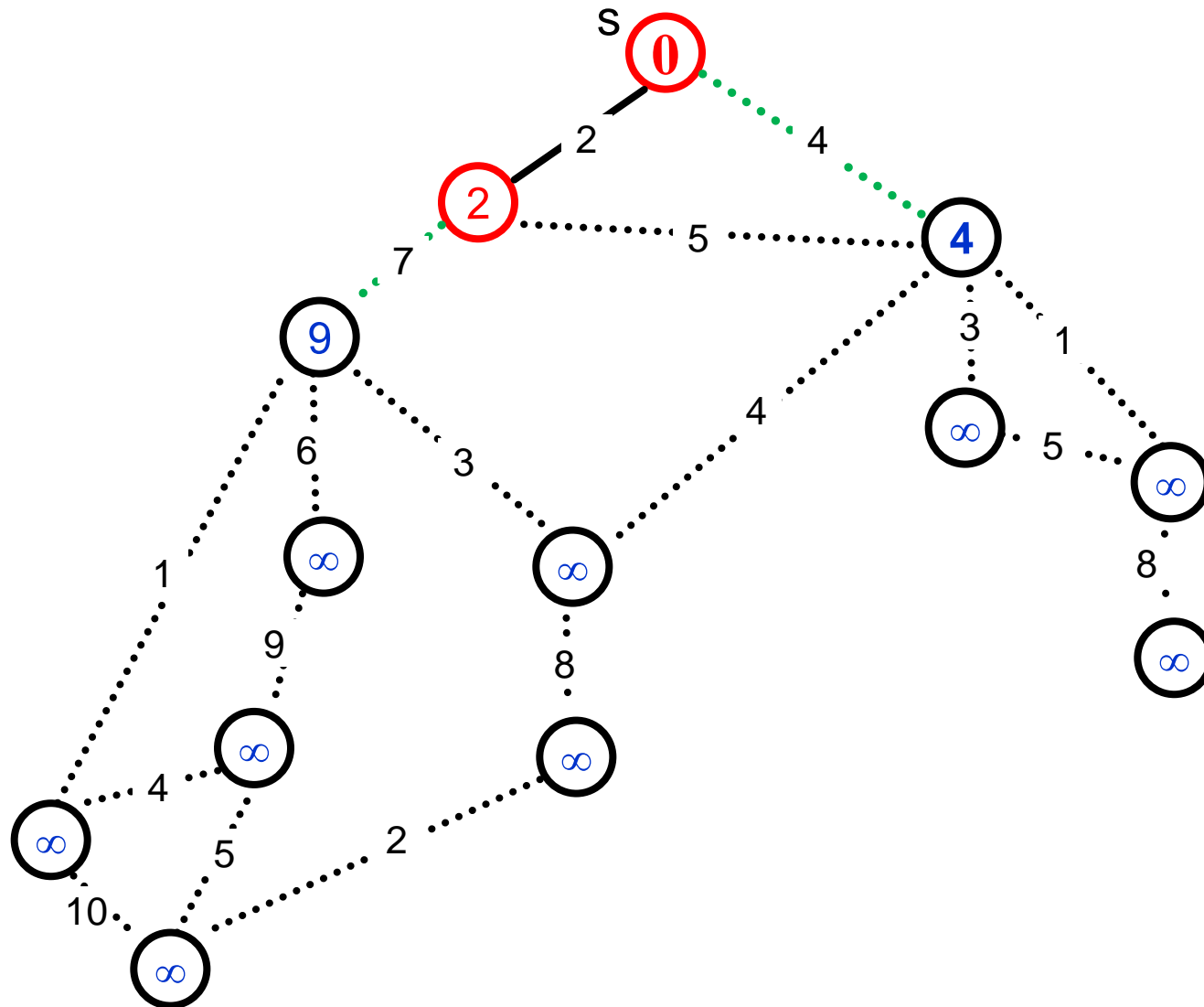
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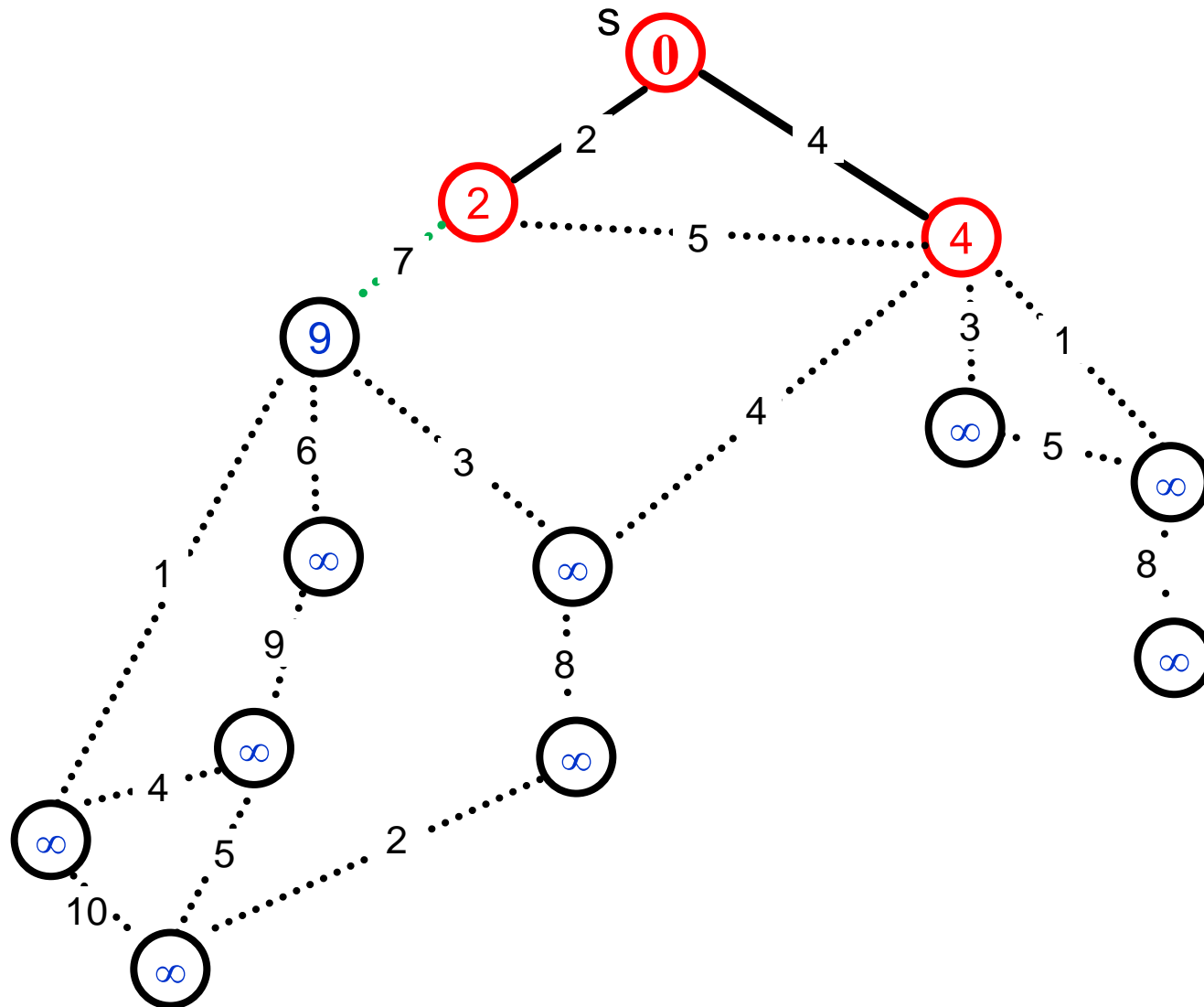
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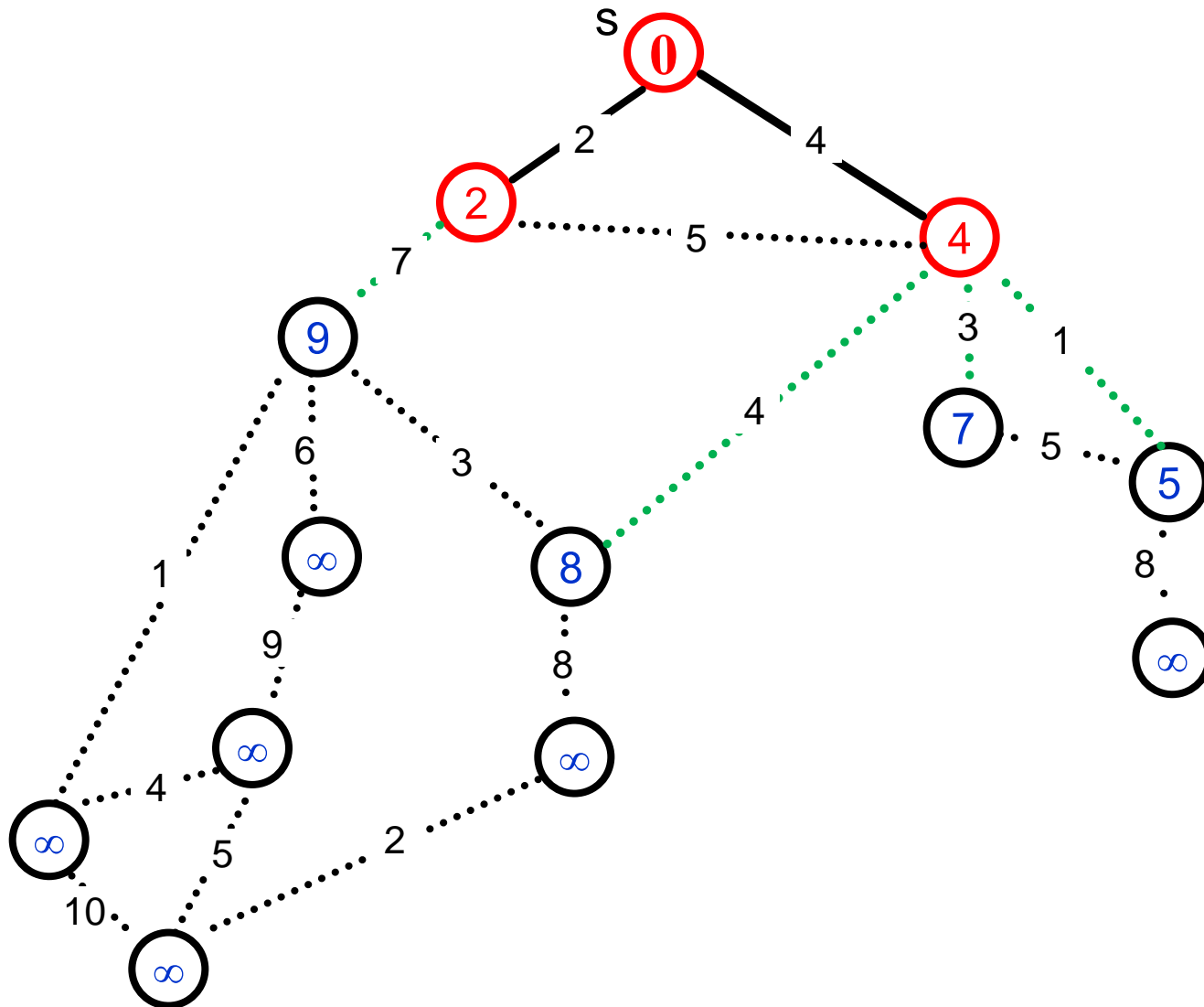
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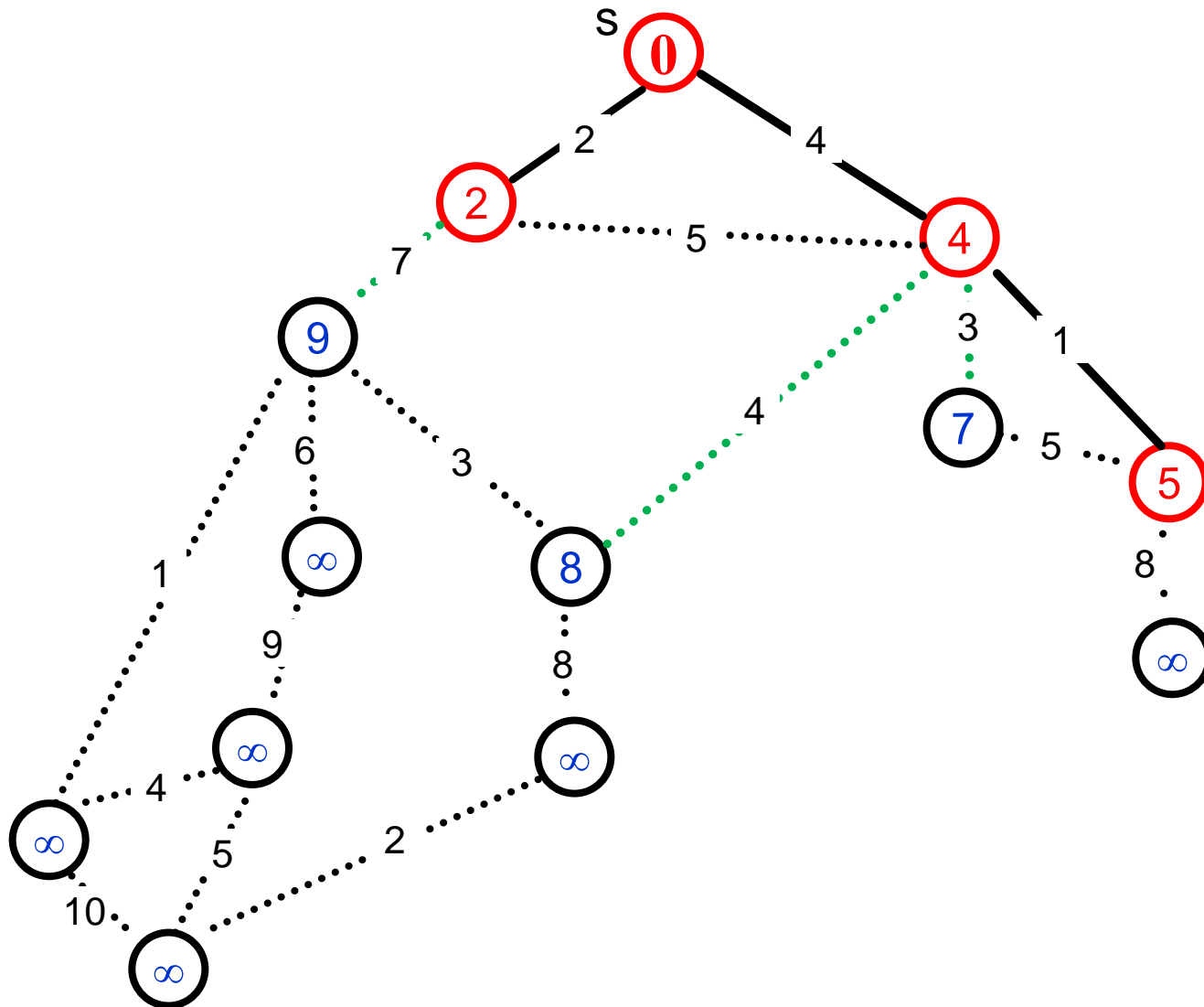


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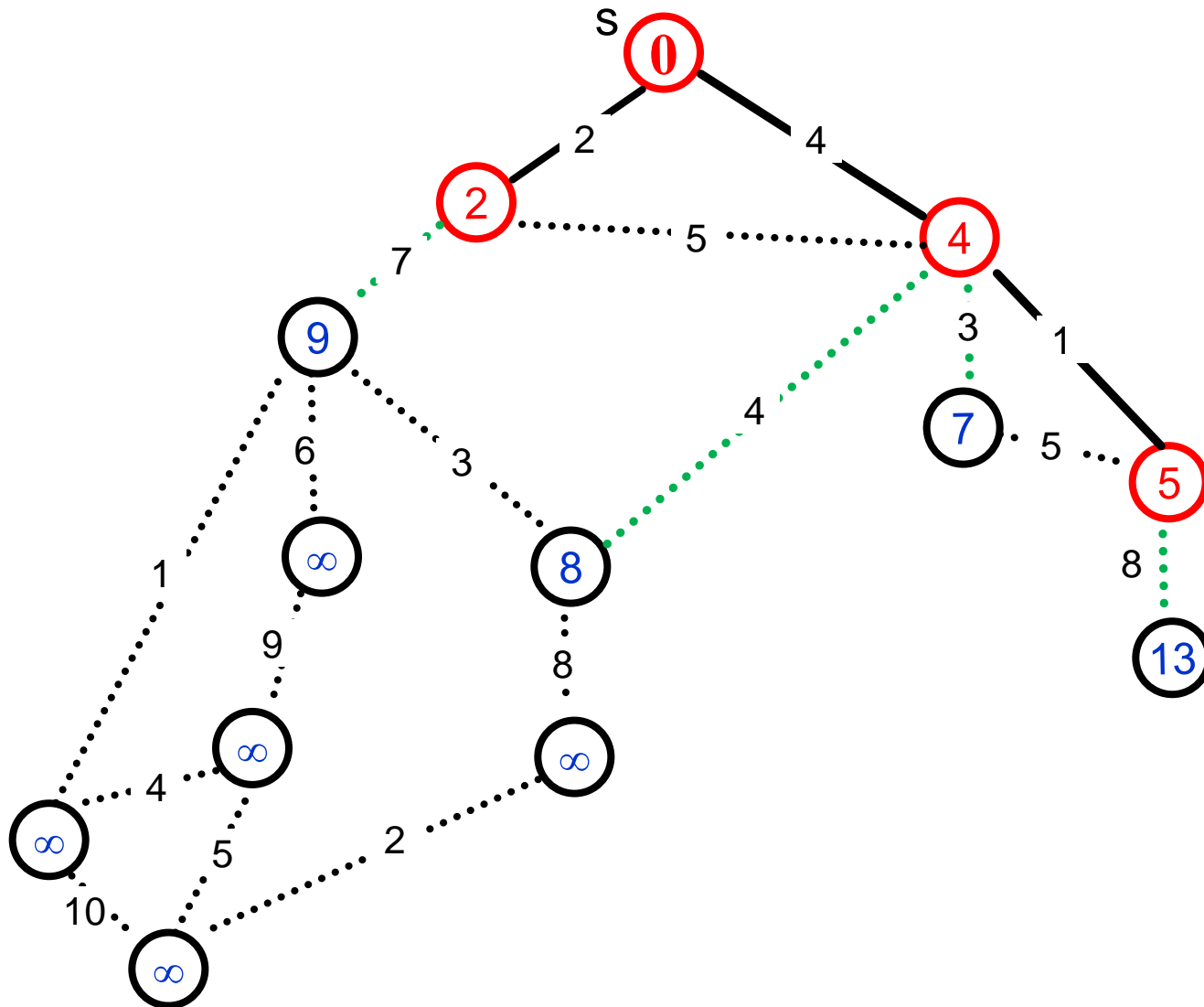




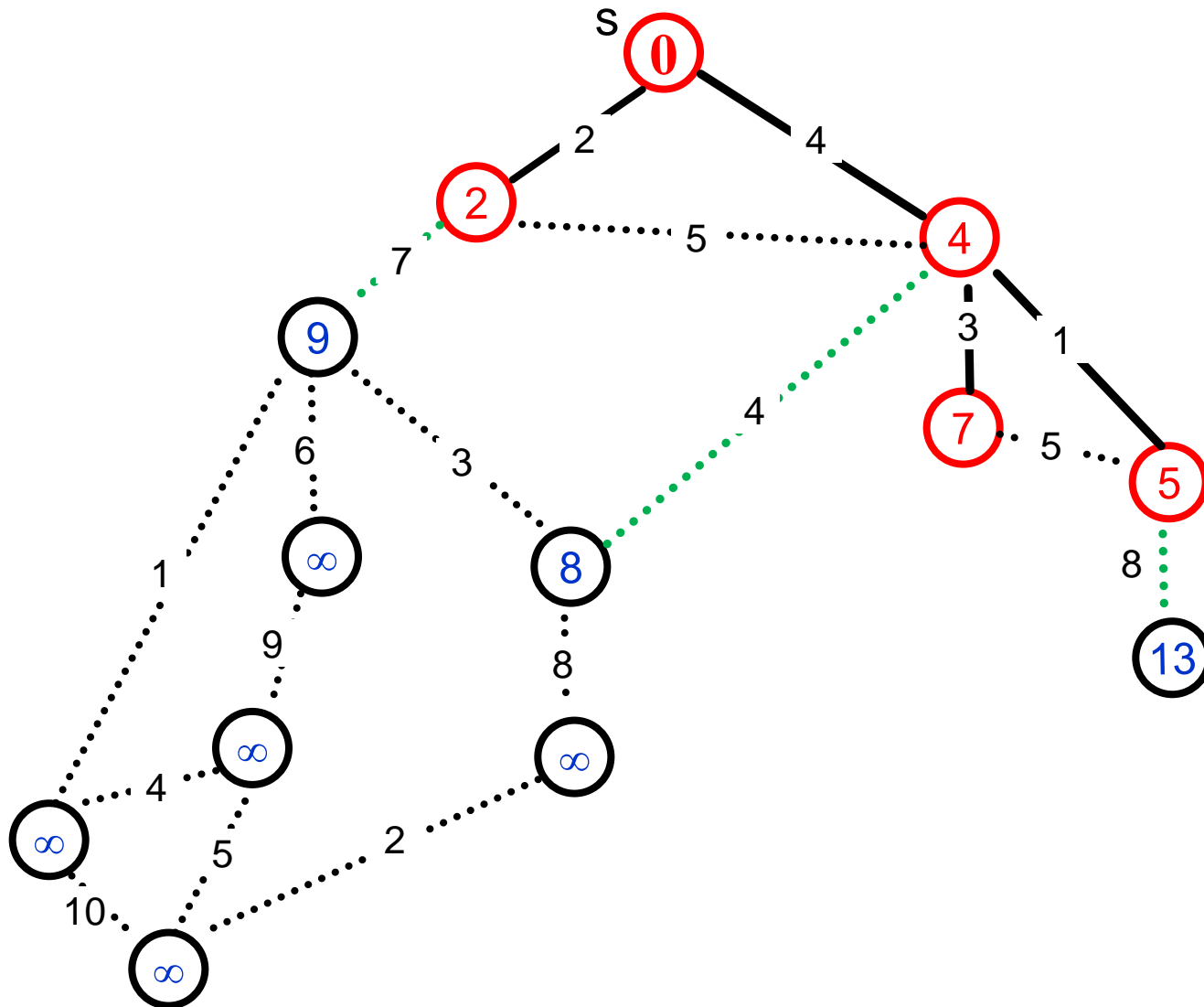
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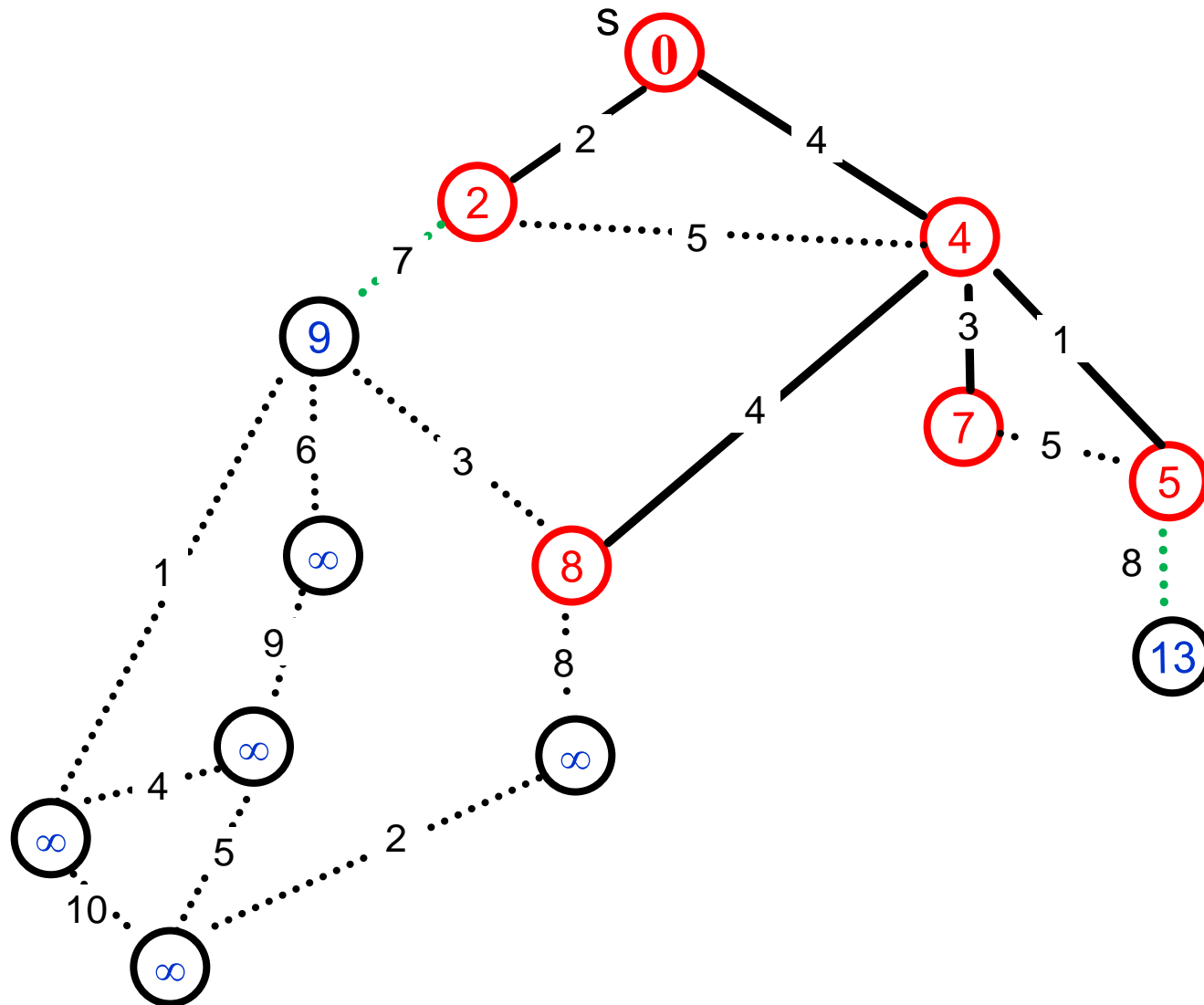
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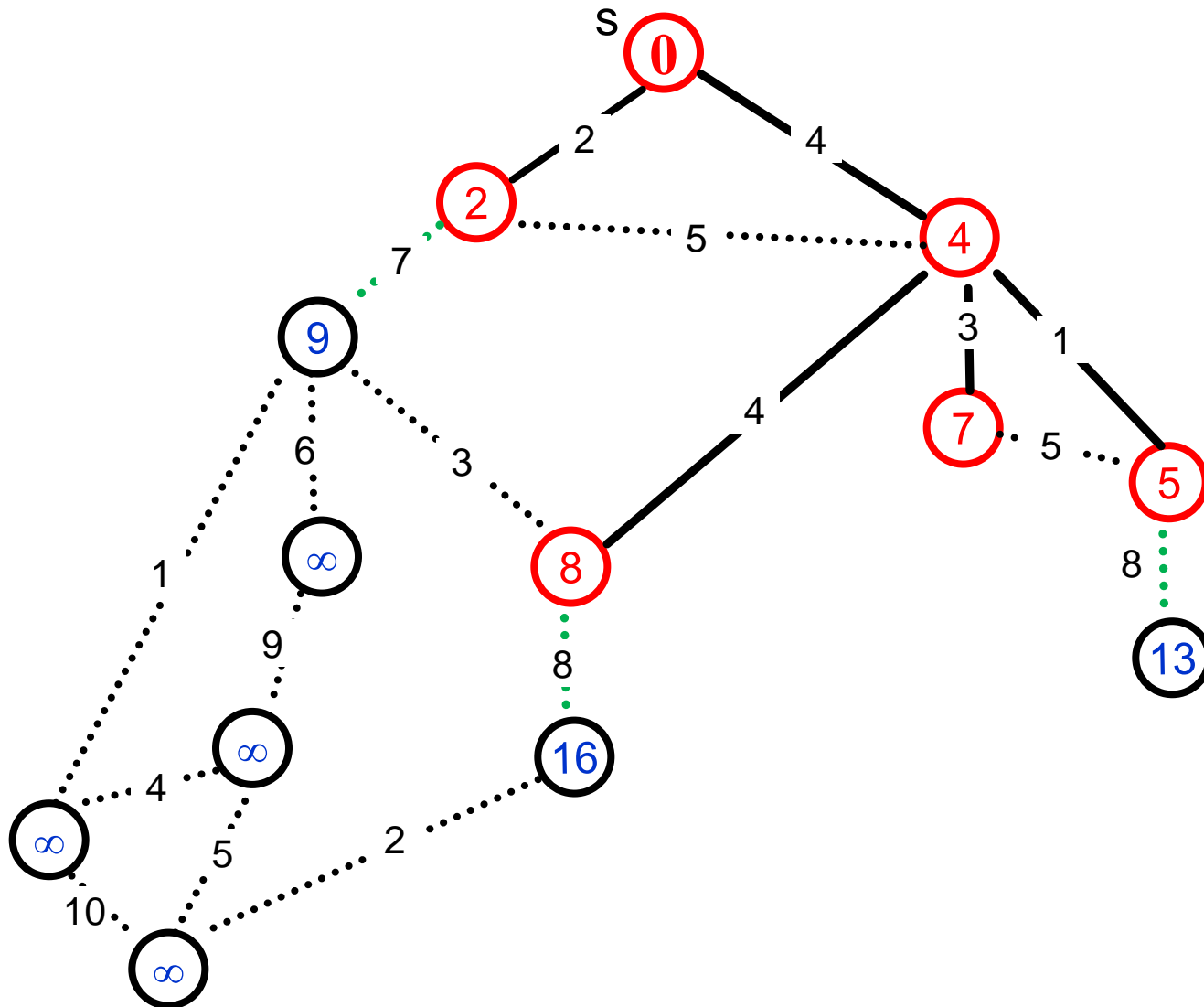
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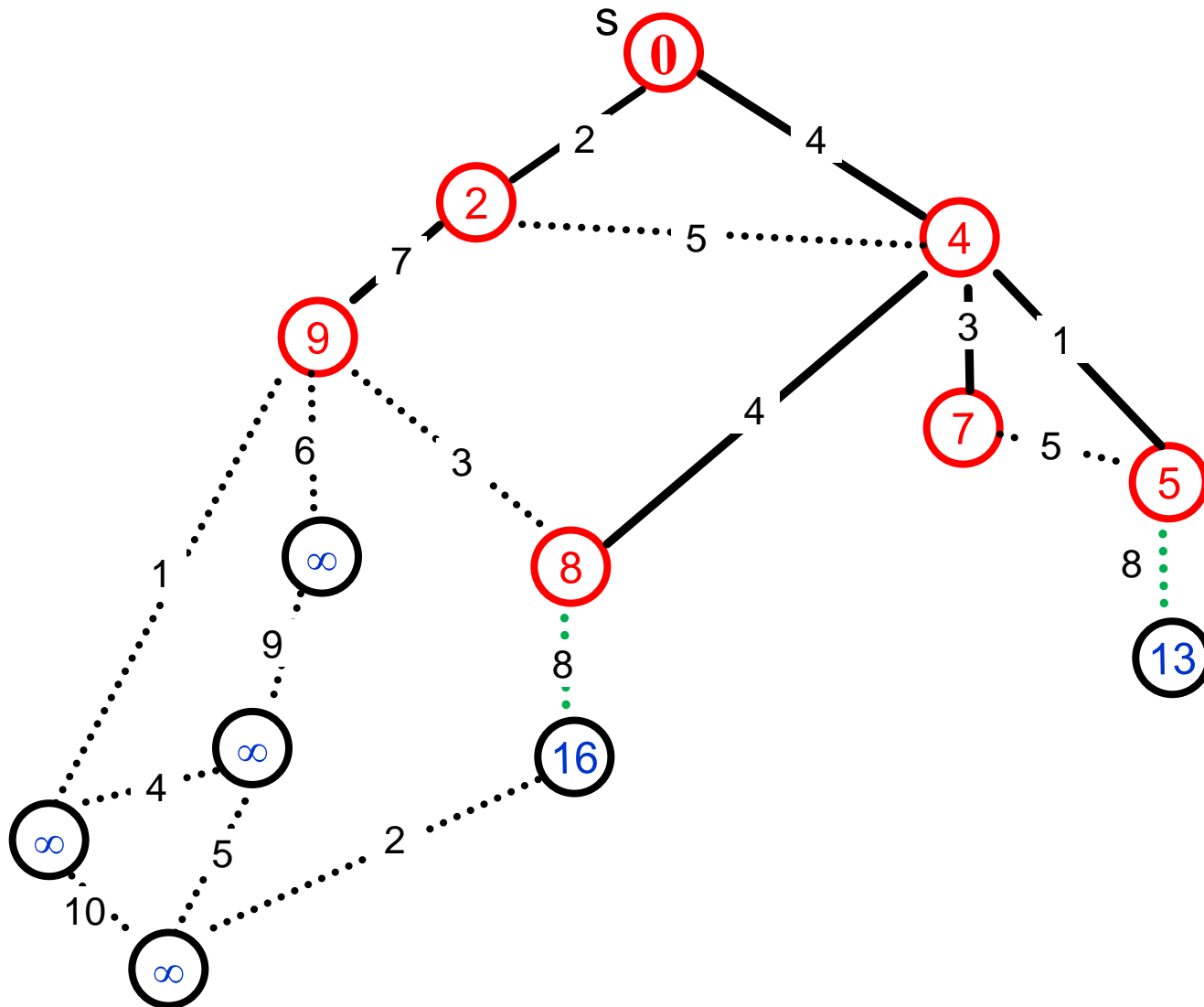
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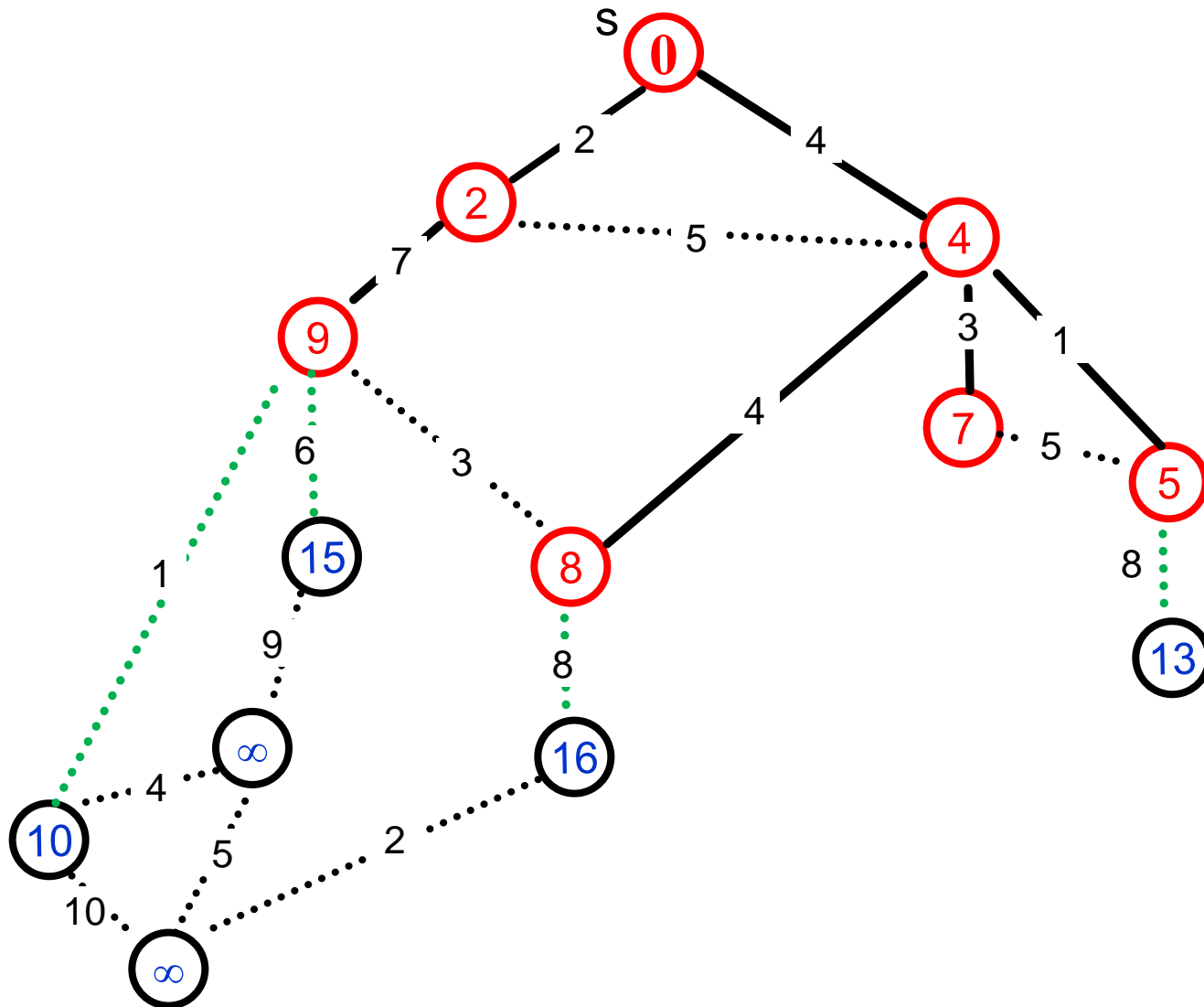
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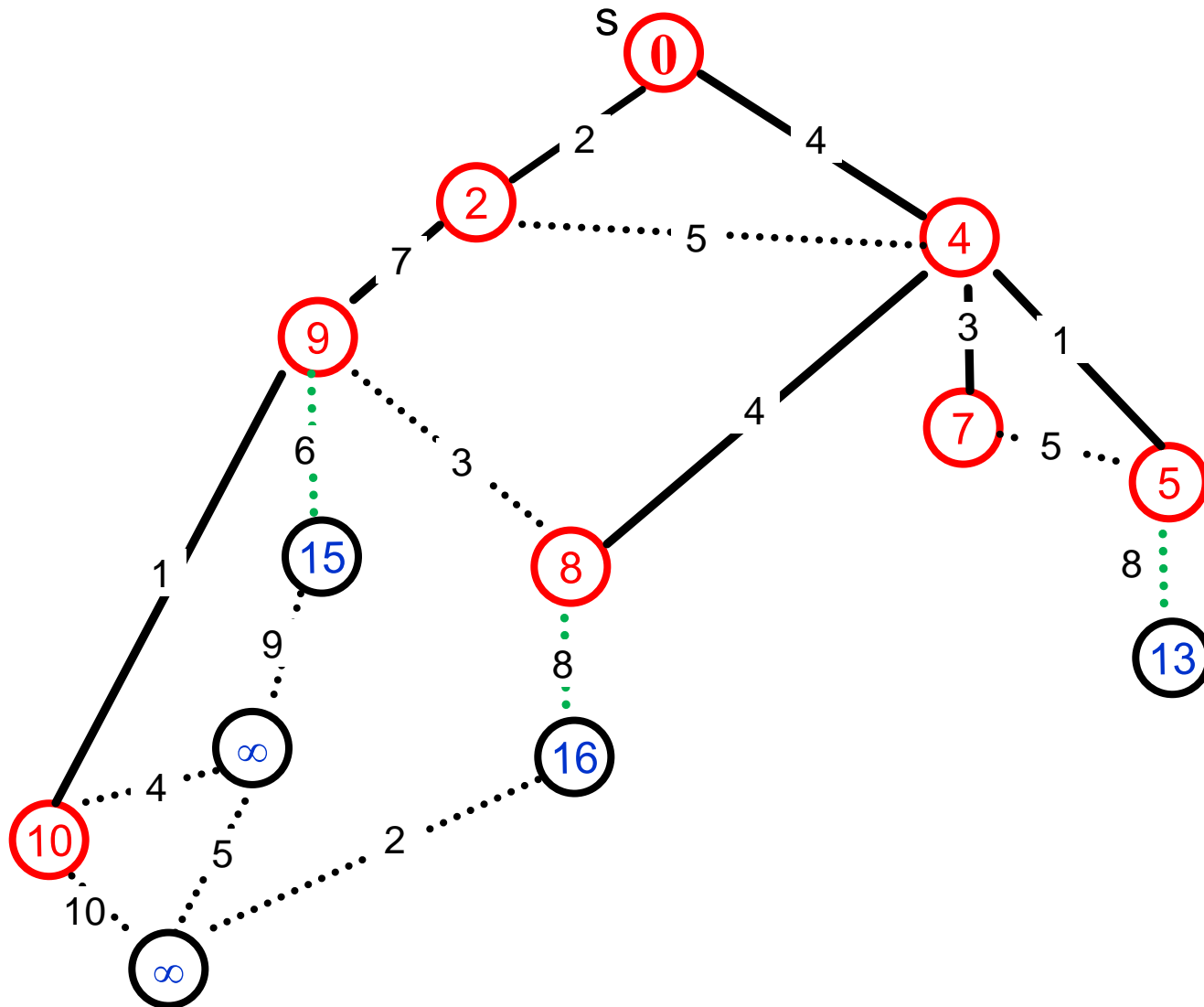
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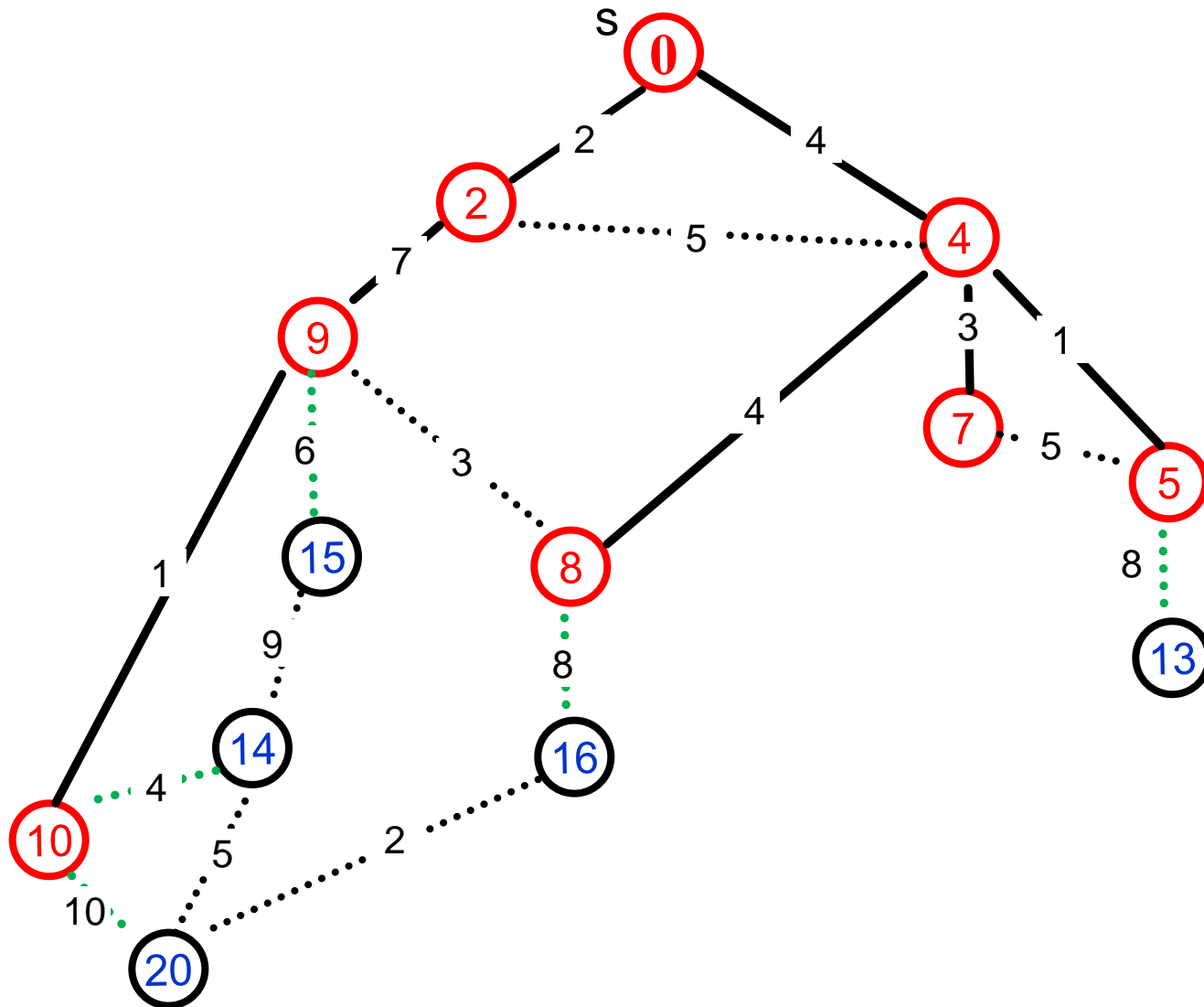


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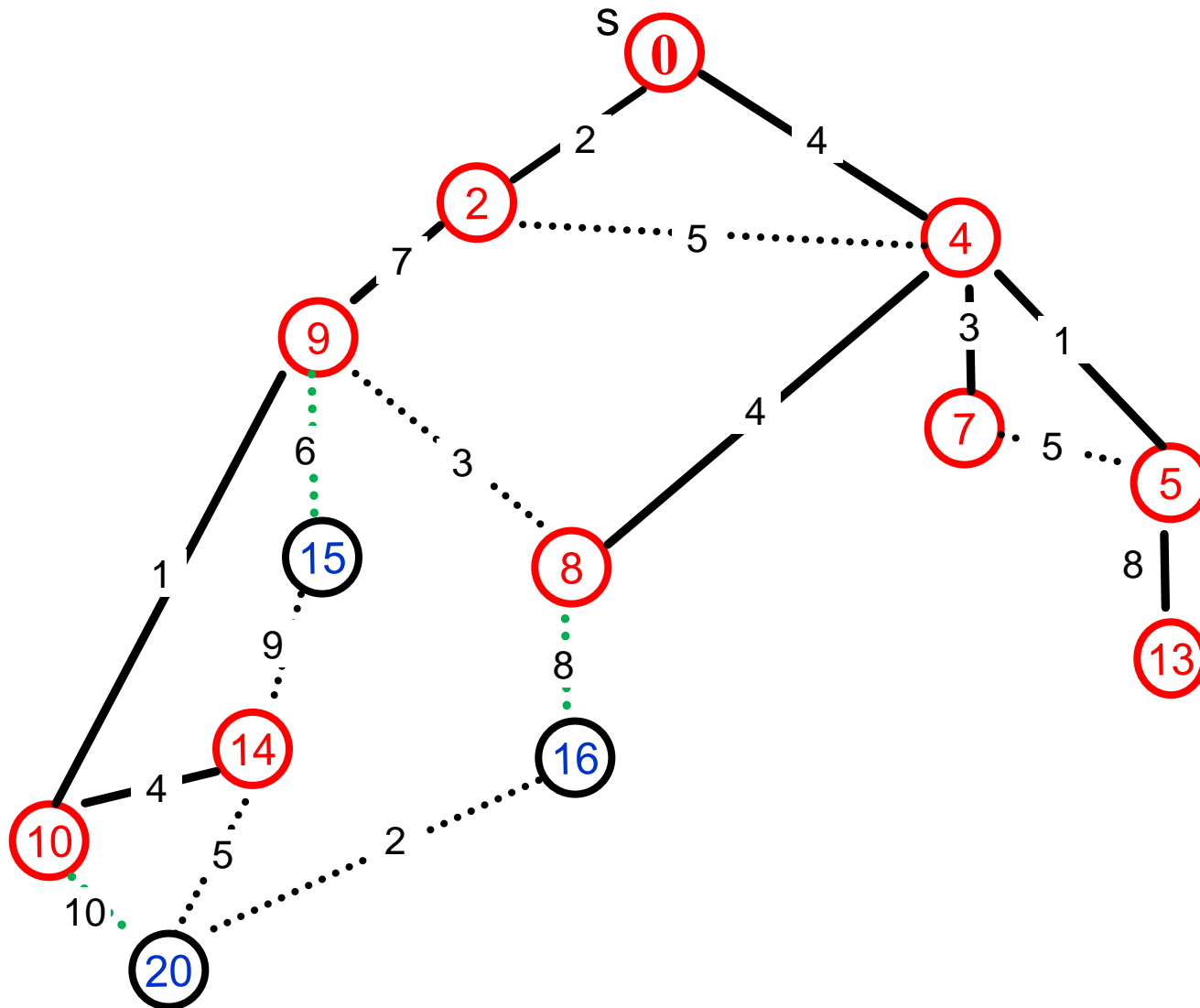




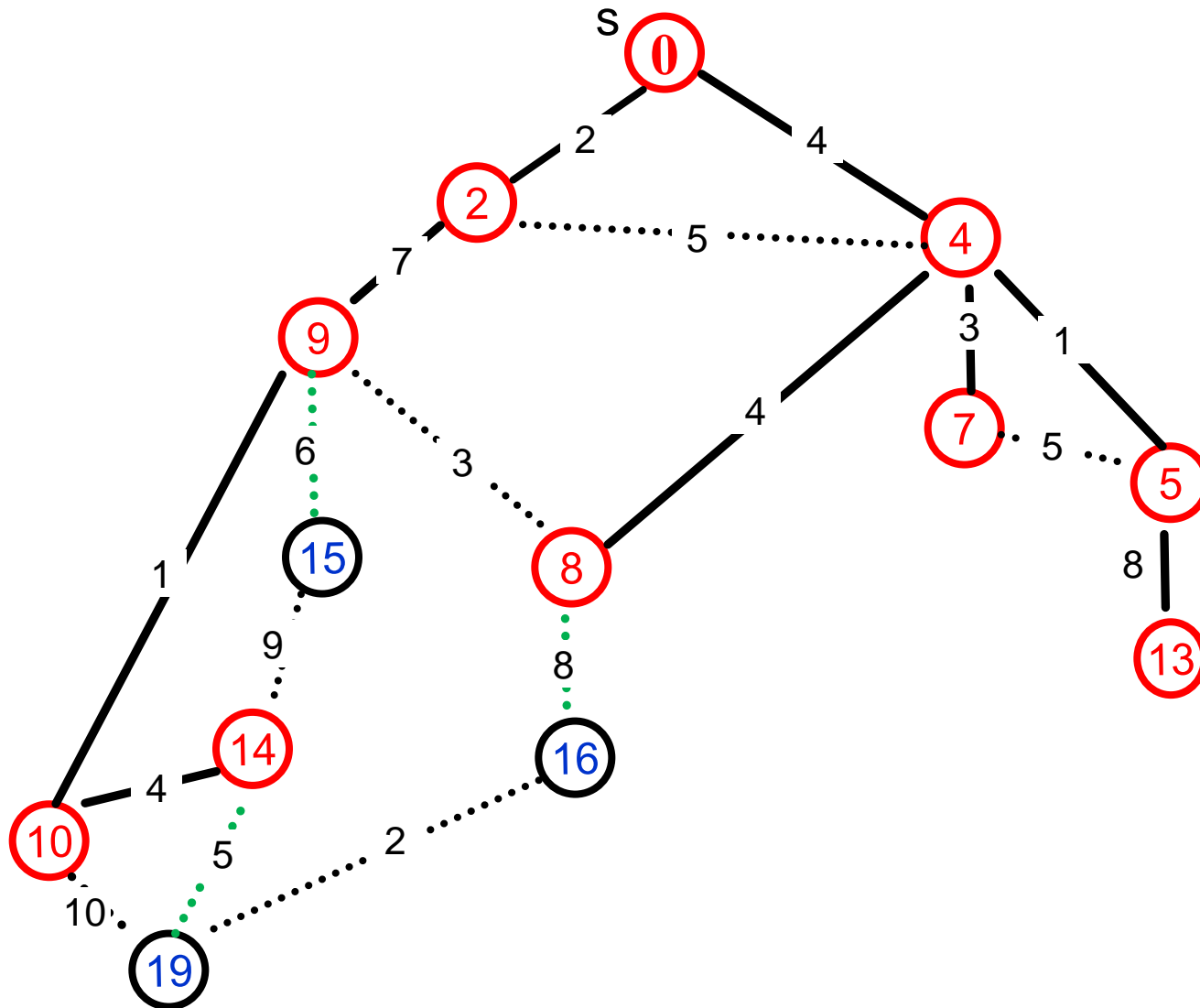
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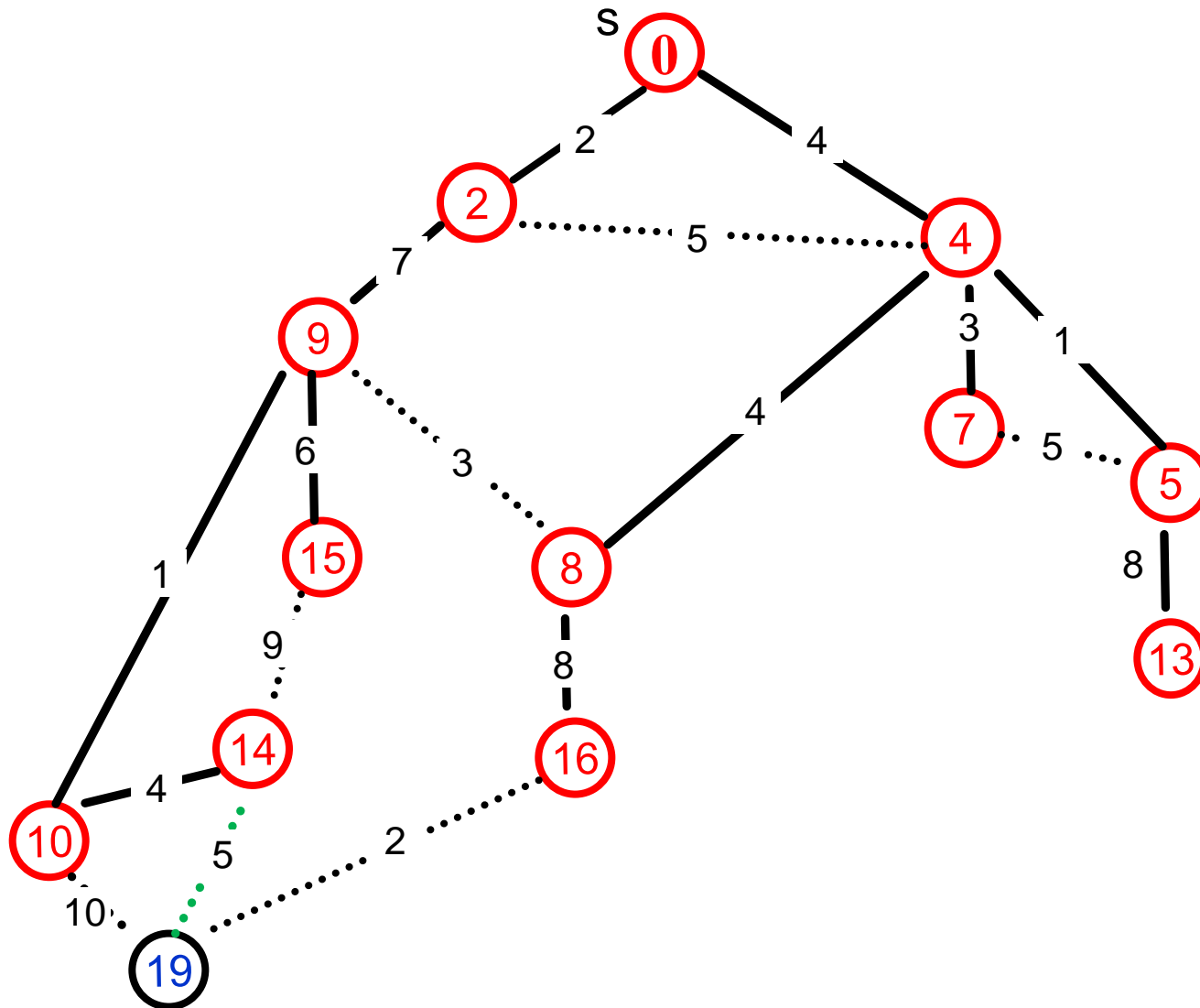
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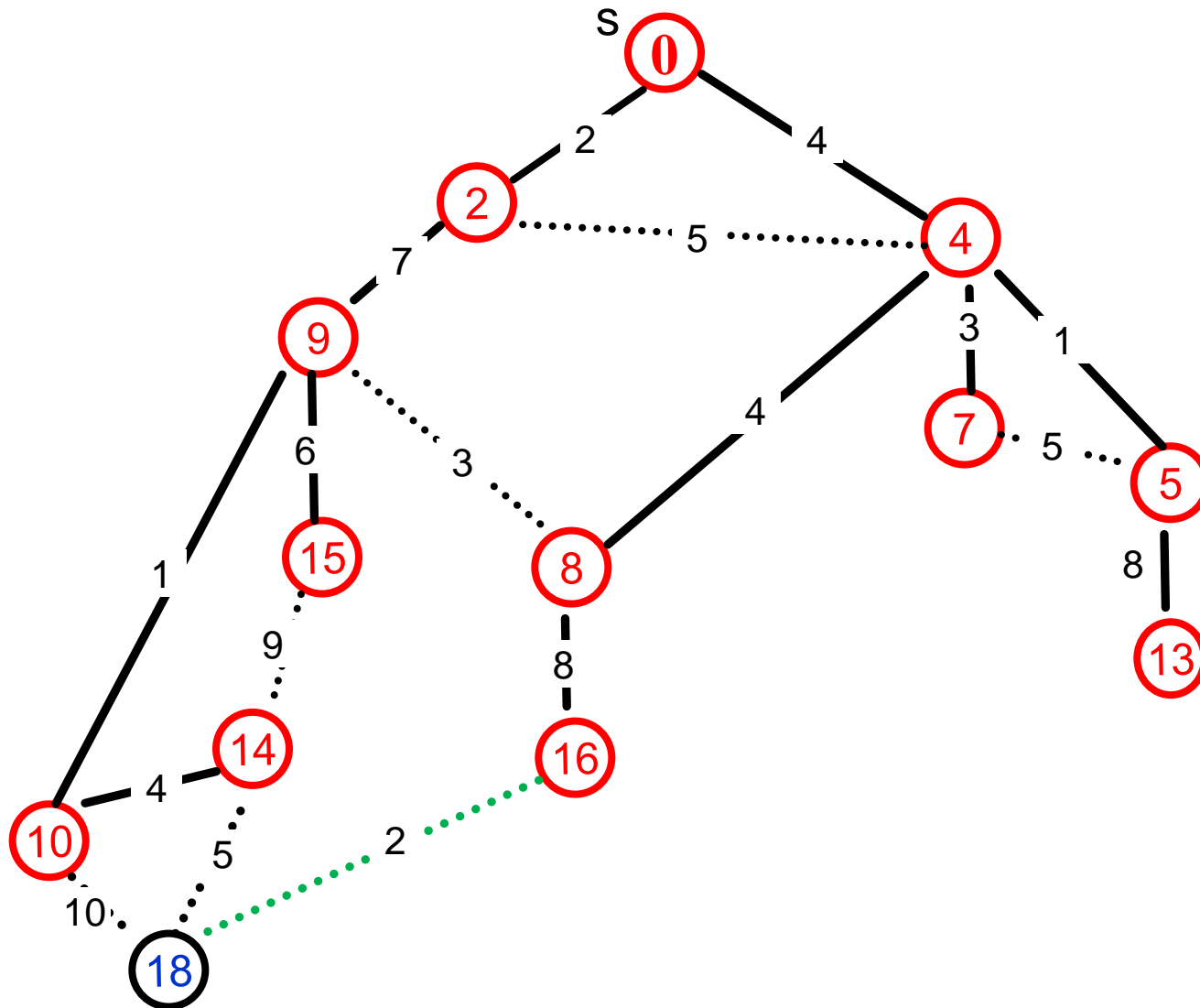
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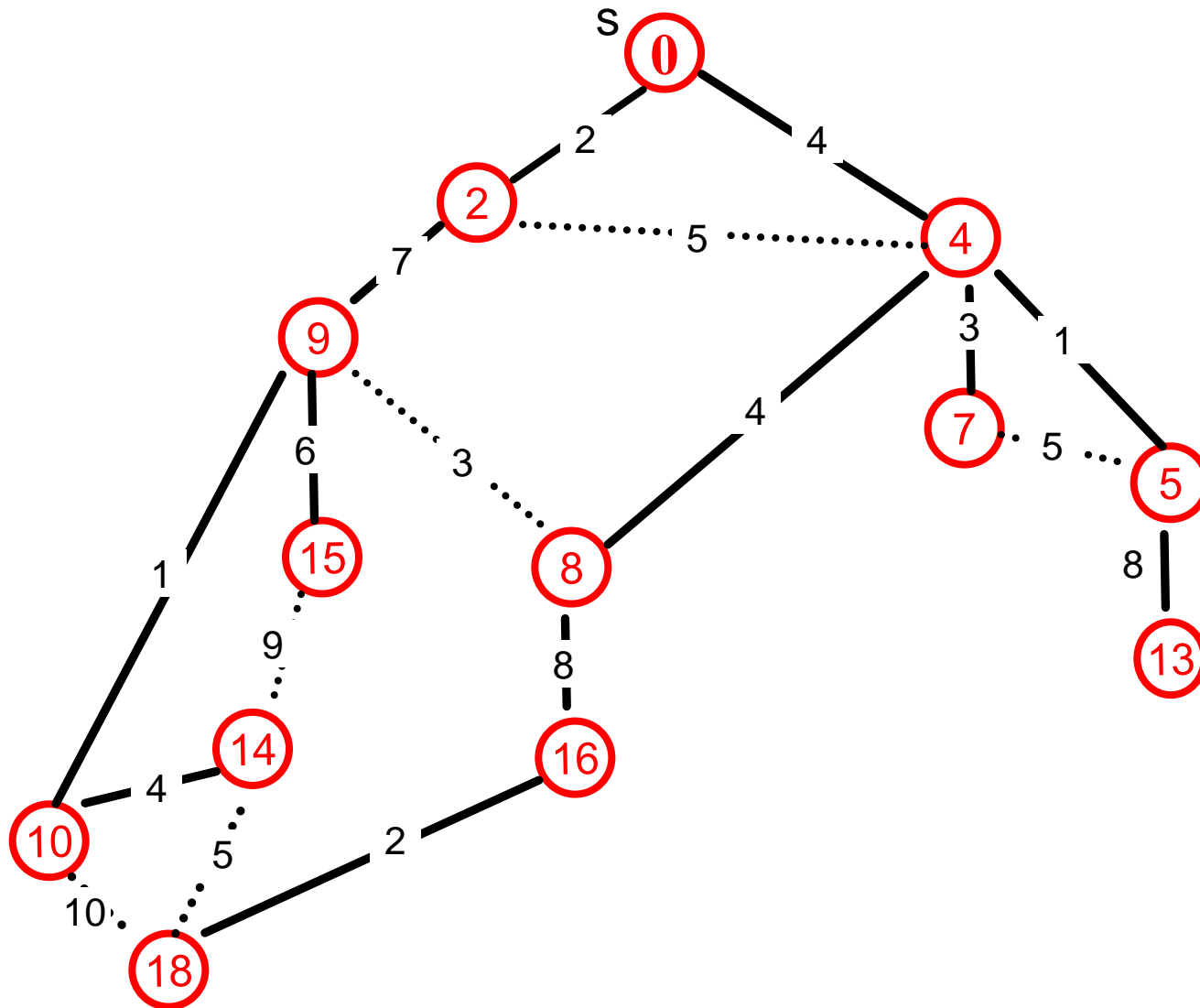
# Dijkstra's Algorithm: Example



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# Disjkstra's Algorithm: Correctness

Prove by induction that throughout the algorithm, for any  $u \in S$ , the path  $P_u$  is the shortest from  $s$  to  $u$ .

**Base Case:** This is always true when  $S = \{s\}$ .

**IH:** Suppose  $|S| = k$  and the claim holds for  $S$

**IS:** Say  $v$  is the  $k+1$ -st vertex that we add to  $S$ . Let  $\{u,v\}$  be last edge on  $P_v$ .

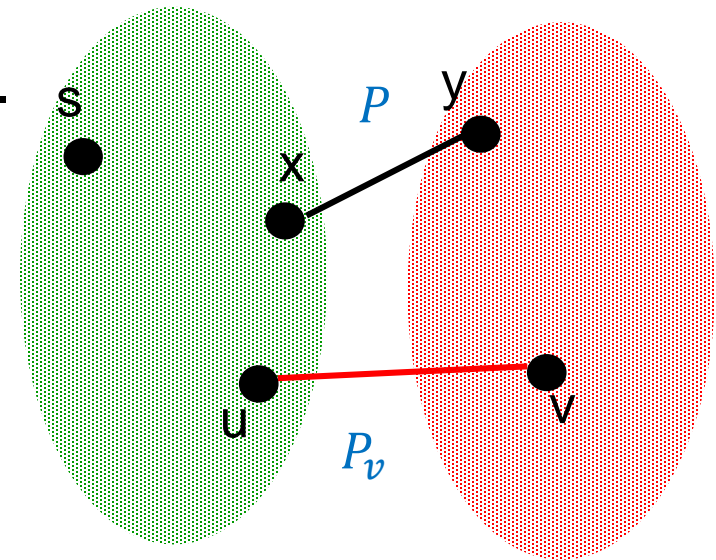
If  $P_v$  is not the shortest path there is a path  $P$  to  $S$  which is shorter.

Consider the **first** time that  $P$  leaves  $S$  (with edge  $\{x,y\}$ ).

$S \rightarrow x$  has weight (at least)  $d(x)$

So,  $c(P) \geq d(x) + c_{x,y} \geq d(v) = c(P_v)$ .

A contradiction.



# Remarks on Dijkstra's Algorithm

- Algorithm also produces a **tree** of shortest paths to  $s$  following Parent links
- Algorithm works on directed graph (with nonnegative weights)
- The algorithm fails with negative edge weights.
  - e.g., some airline tickets

Why does it fail?

- Dijkstra's algorithm is similar to BFS:
  - Substitute every edge with  $c_e = k$  with a path of length  $k$ , then run BFS.



# Implementing Dijkstra's Algorithm

**Priority Queue:** Elements each with an associated key Operations

- Insert
- Find-min
  - Return the element with the smallest key
- Delete-min
  - Return the element with the smallest key and delete it from the data structure
- Decrease-key
  - Decrease the key value of some element

Implementations

Arrays:

- $O(n)$  time find/delete-min,
- $O(1)$  time insert/decrease key

Binary Heaps:

- $O(\log n)$  time insert/decrease-key/delete-min,
- $O(1)$  time find-min

# Dijkstra's Algorithm

Runs in  $O((n+m)\log n)$ .

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    }  
}
```

$O(n)$  of delete min,  
each in  $O(\log n)$

$O(m)$  of decrease key,  
each runs in  $O(\log n)$

# Summary (Greedy Algorithms)

- **Greedy Stays Ahead:** Interval Scheduling, Dijkstra's algorithm
- **Structural:** Interval Partitioning
- **Exchange Arguments:** MST, Kruskal's Algorithm, Prim's Algorithm
- **Data Structures:** Union Find, Heap

# Divide and Conquer Approach

# Divide and Conquer

Similar to algorithm design by induction, we reduce a problem to several subproblems.

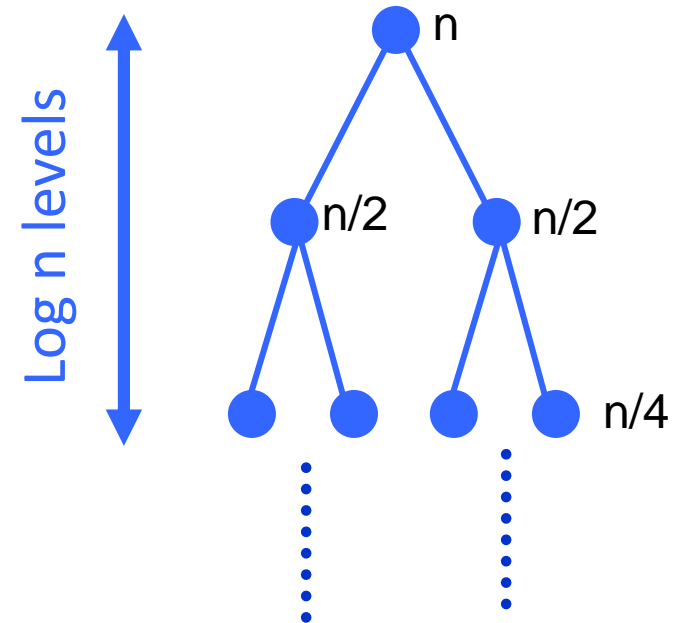
Typically, each sub-problem is **at most a constant fraction** of the size of the original problem

Recursively solve each subproblem

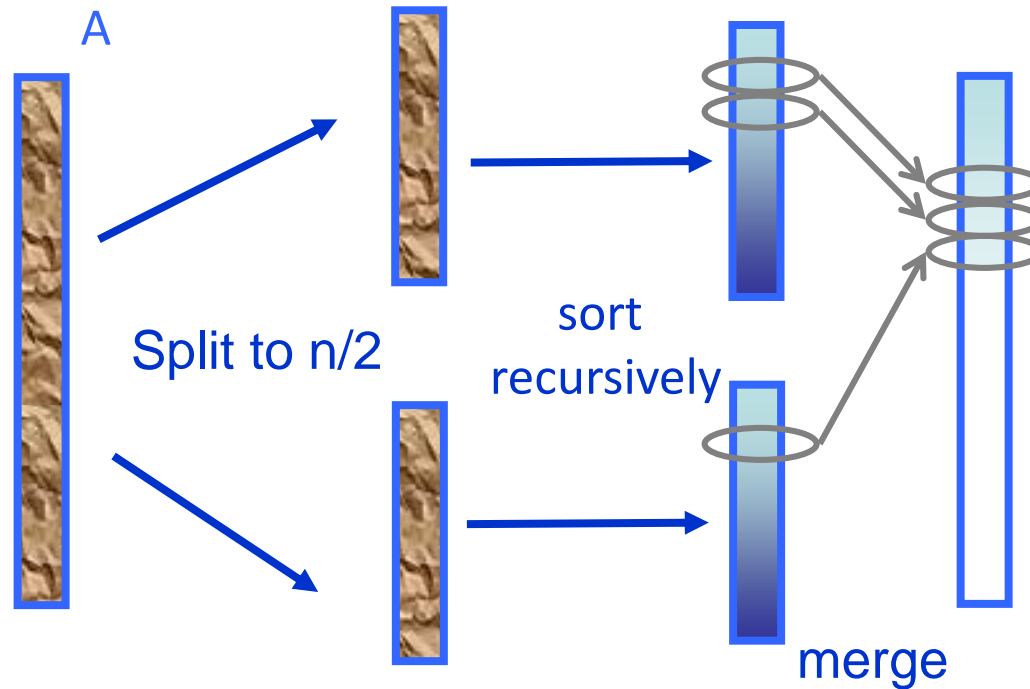
Merge the solutions

## Examples:

- Mergesort, Binary Search, Strassen's Algorithm,



# A Classical Example: Merge Sort



# Why Balanced Partitioning?

An alternative "divide & conquer" algorithm:

- Split into  $n-1$  and  $1$
- Sort each sub problem
- Merge them

## Runtime

$$T(n) = T(n - 1) + T(1) + n$$

## Solution:

$$\begin{aligned} T(n) &= n + T(n - 1) + T(1) \\ &= n + n - 1 + T(n - 2) \\ &= n + n - 1 + n - 2 + T(n - 3) \\ &= n + n - 1 + n - 2 + \cdots + 1 = O(n^2) \end{aligned}$$

# D&C: The Key Idea

Suppose we've already invented Bubble-Sort, and we know it takes  $n^2$

Try **just one level** of divide & conquer:

Bubble-Sort(first  $n/2$  elements)

Bubble-Sort(last  $n/2$  elements)

Merge results

Time:  $2 (n/2)^2 + n = n^2/2 + n \ll n^2$

Almost twice as fast!





# D&C approach

- “the more dividing and conquering, the better”
  - Two levels of D&C would be almost 4 times faster, 3 levels almost 8, etc., even though overhead is growing.
  - Best is usually full recursion **down to a small constant** size (balancing "work" vs "overhead").

In the limit: you’ve just rediscovered mergesort!

- Even unbalanced partitioning is good, but less good
  - Bubble-sort improved with a 0.1/0.9 split:

$$(.1n)^2 + (.9n)^2 + n = .82n^2 + n$$

The 18% savings compounds significantly if you carry recursion to more levels, actually giving  $O(n \log n)$ , but with a bigger constant.

- This is why Quicksort with random splitter is good – badly unbalanced splits are rare, and not instantly fatal.

# Finding the Root of a Function

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Given a continuous function  $f$  and two points  $a < b$  such that

$$f(a) \leq 0$$

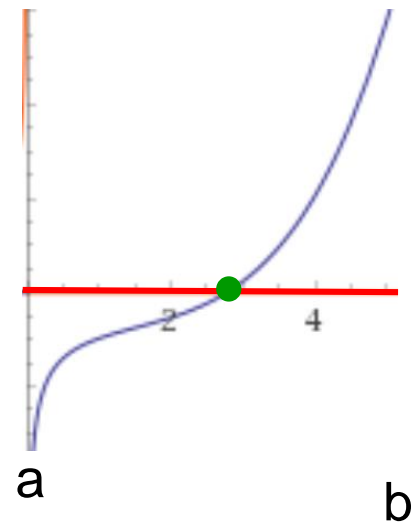
$$f(b) \geq 0$$

Find an approximate root of  $f$  (a point  $c$  where  $f(c) = 0$ ).

$f$  has a root in  $[a, b]$  by  
intermediate value theorem

Note that roots of  $f$  may be **irrational**,  
So, we want to approximate  
the root with an arbitrary precision!

$$f(x) = \sin(x) - \frac{100}{\sqrt{x}} + x^4$$



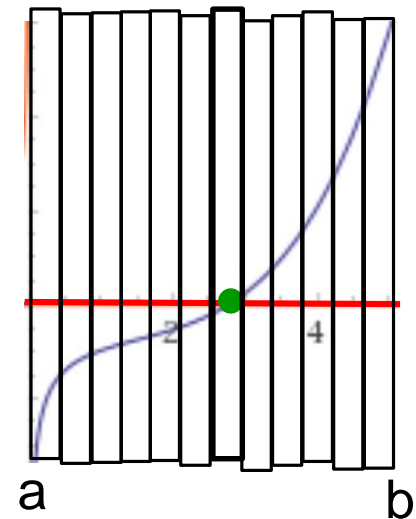
# A Naïve Approach

Suppose we want  $\epsilon$  approximation to a root.

Divide  $[a,b]$  into  $n = \frac{b-a}{\epsilon}$  intervals. For each interval check  
$$f(x) \leq 0, f(x + \epsilon) \geq 0$$

This runs in time  $O(n) = O(\frac{b-a}{\epsilon})$

Can we do faster?



# D&C Approach (Based on Binary Search)

**Bisection**(a,b,  $\epsilon$ )

if  $(b - a) < \epsilon$  then

return (a)

else

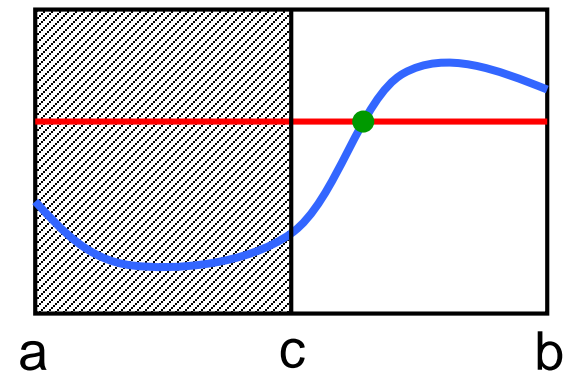
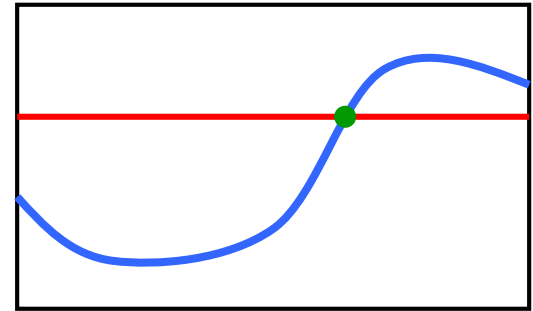
$m \leftarrow (a + b)/2$

if  $f(m) \leq 0$  then

return(Bisection(c, b,  $\epsilon$ ))

else

return(Bisection(a, c,  $\epsilon$ ))



# Time Analysis

Let  $n = \frac{a-b}{\epsilon}$

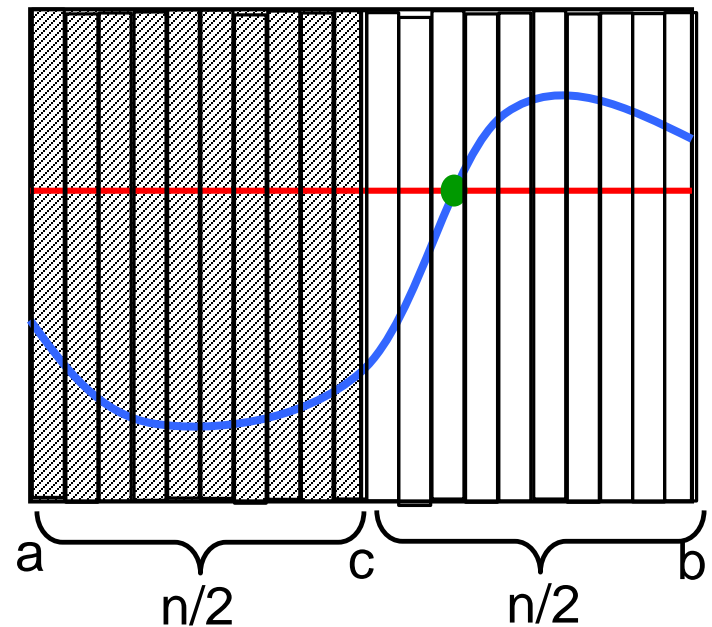
And  $c = (a + b)/2$

Always half of the intervals lie to the left and half lie to the right of  $c$

So,

$$T(n) = T\left(\frac{n}{2}\right) + O(1)$$

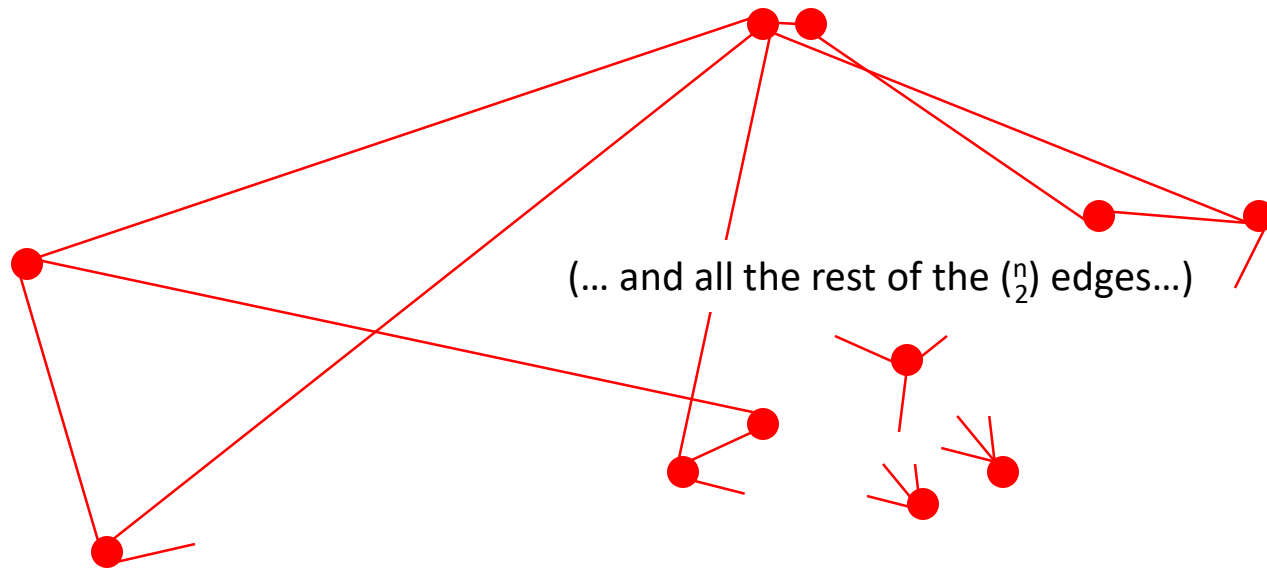
i.e.,  $T(n) = O(\log n) = O\left(\log \frac{a-b}{\epsilon}\right)$



# Finding the Closest Pair of Points

# Closest Pair of Points (non geometric)

Given  $n$  points and **arbitrary** distances between them, find the closest pair. (E.g., think of distance as airfare – definitely not Euclidean distance!)



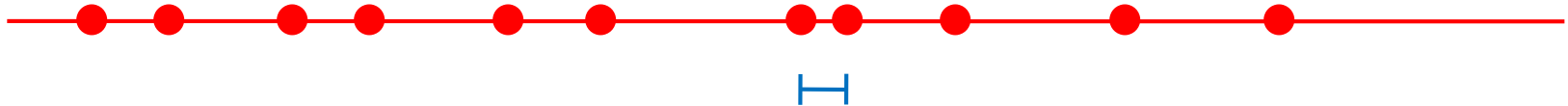
*Must* look at all  $n$  choose 2 pairwise distances, else any one you didn't check might be the shortest.

i.e., you have to read the whole input



# Closest Pair of Points (1-dimension)

Given  $n$  points on the real line, find the closest pair



**Fact:** Closest pair is **adjacent** in ordered list

So, first sort, then scan adjacent pairs.

Time  $O(n \log n)$  to sort, if needed, Plus  $O(n)$  to scan adjacent pairs

**Key point:** do *not* need to calc distances between all pairs: exploit geometry + ordering

# Closest Pair of Points (2-dimensions)

Given  $n$  points in the plane, find a pair with smallest Euclidean distance between them.

## Fundamental geometric primitive.

Graphics, computer vision, geographic information systems, molecular modeling, air traffic control.

Special case of nearest neighbor, Euclidean MST, Voronoi.

**Brute force:** Check all pairs of points  $p$  and  $q$  with  $\Theta(n^2)$  time.

**Assumption:** No two points have same  $x$  coordinate.

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# A Divide and Conquer Alg

**Divide:** draw vertical line  $L$  with  $\approx n/2$  points on each side.

**Conquer:** find closest pair on each side, recursively.

**Combine** to find closest pair overall

← seems like  $\Theta(n^2)$  ?

Return best solutions

