

# DSC 40B

## Theoretical Foundations II

Dijkstra's Algorithm

# Shortest Path Algorithms

- ▶ **Bellman-Ford** and **Dijkstra's** are shortest path algorithms:

INPUT: weighted graph, source vertex  $s$ .

OUTPUT: shortest paths from  $s$  to every other node.

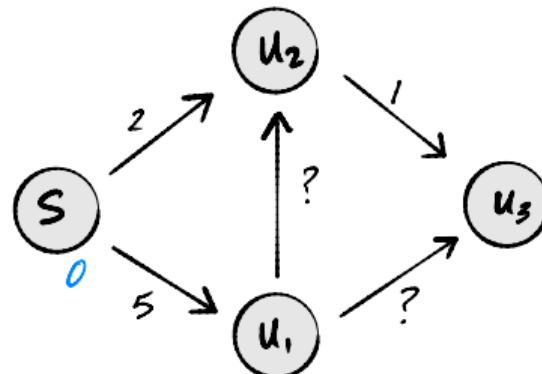
- ▶ Both work by:
  - ▶ keeping estimates of shortest path distances;
  - ▶ iteratively **updating** estimates until they're correct.

# Shortest Path Algorithms

- ▶ We saw Bellman-Ford last time; takes time  $\Theta(VE)$ .
- ▶ Dijkstra's will be faster, but can't handle negative weights.

# Dijkstra's Algorithm

- ▶ On every iteration, Bellman-Ford updates all edges – many don't need to be updated.
- ▶ If we **assume** all edge weights are positive, we can rule out some paths immediately:



# Dijkstra's Idea

- ▶ Keep track of set  $C$  of correct nodes.
- ▶ At every step, add node outside of  $C$  with smallest estimated distance; update its neighbors.

# Outline of Dijkstra's Algorithm

```
def dijkstra(graph, weights, source):
    est = {node: float('inf') for node in graph.nodes}
    est[source] = 0
    pred = {node: None for node in graph.nodes}

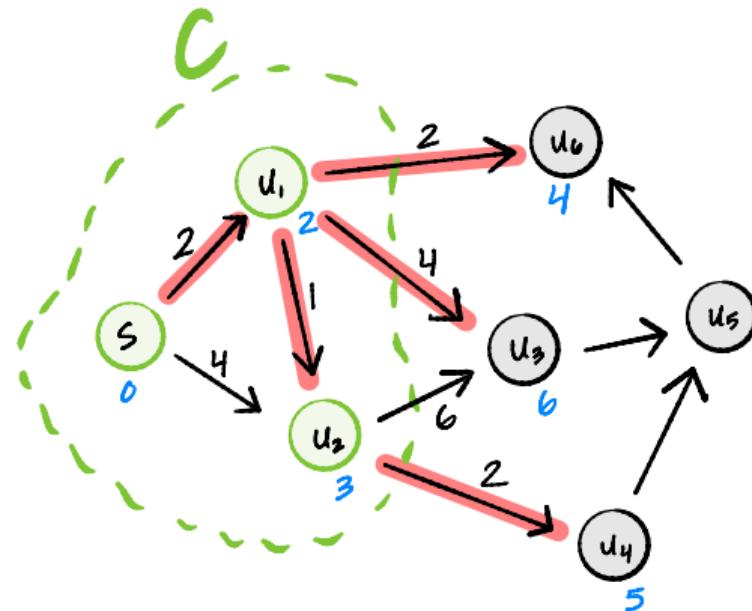
    # empty set
    C = set()

    # while there are nodes still outside of C
        # find node u outside of C with smallest
        # estimated distance
    C.add(u)
    for v in graph.neighbors(u):
        update(u, v, weights, est, pred)

return est, pred
```

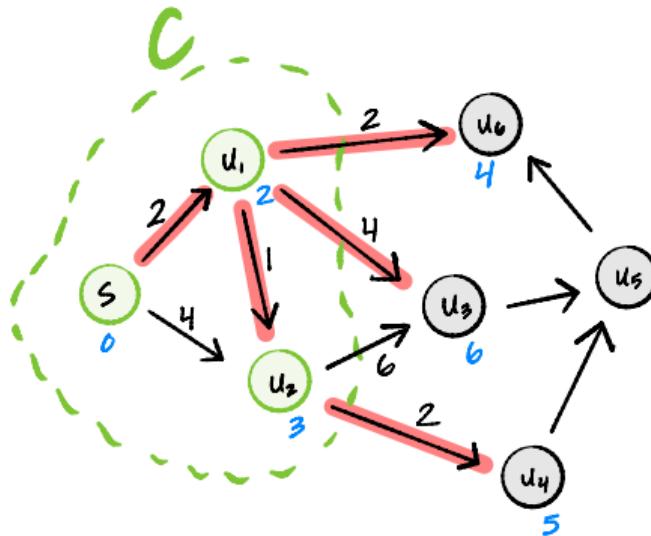
# Proof Idea

- Claim: at beginning of any iteration of Dijkstra's, if  $u$  is node  $\notin C$  with smallest estimated distance, the shortest path to  $u$  has been correctly discovered.



# Proof Idea

- ▶ Let  $u$  be node outside of  $C$  for which  $\text{est}[u]$  is smallest.
- ▶ We've discovered a path from  $s$  to  $u$  of length  $\text{est}[u]$ .
- ▶ Any path from  $s$  to  $u$  has to exit  $C$  somewhere.
- ▶ Any path from  $s$  to  $u$  will cost at least  $\text{est}[u]$  just to exit  $C$ .



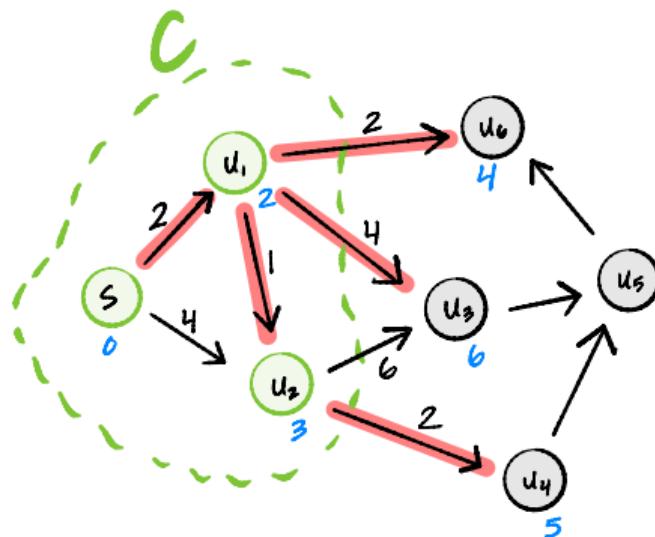
# DSC 40B

## Theoretical Foundations II

Analysis

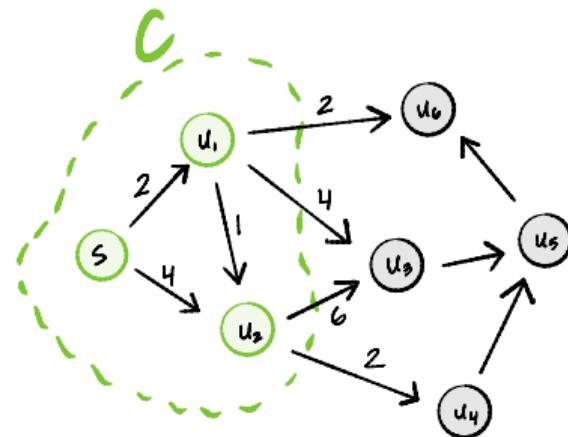
# Proof Idea

- ▶ Let  $u$  be node outside of  $C$  for which  $\text{est}[u]$  is smallest.
- ▶ We've discovered a path from  $s$  to  $u$  of length  $\text{est}[u]$ .
- ▶ Any path from  $s$  to  $u$  has to exit  $C$  somewhere.
- ▶ Any path from  $s$  to  $u$  will cost at least  $\text{est}[u]$  just to exit  $C$ .



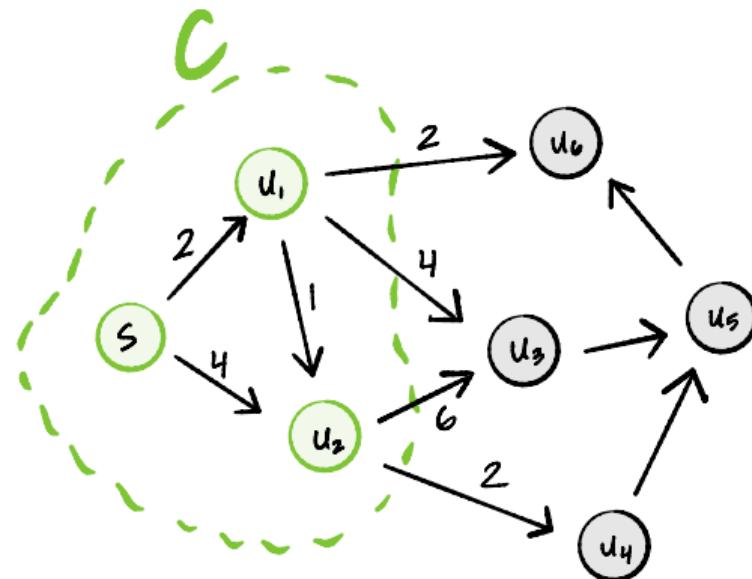
# Exit Paths

- ▶ An **exit path from  $s$  through  $C$**  is a path for which:
  - ▶ the first node is  $s$ ;
  - ▶ the last node (a.k.a., the **exit node**) is **not** in  $C$ ;
  - ▶ all other nodes **are** in  $C$ .
- ▶ Example:



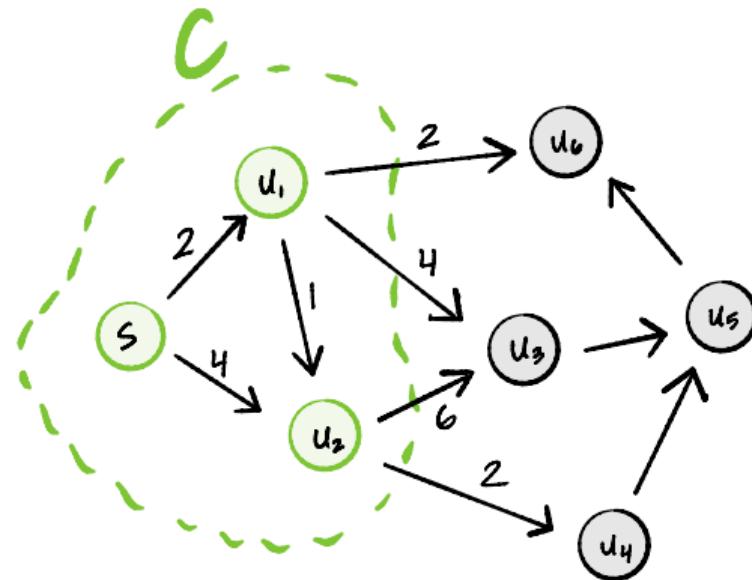
# Exit Paths

- True or False: this is an exit path from  $s$  through  $C$ .



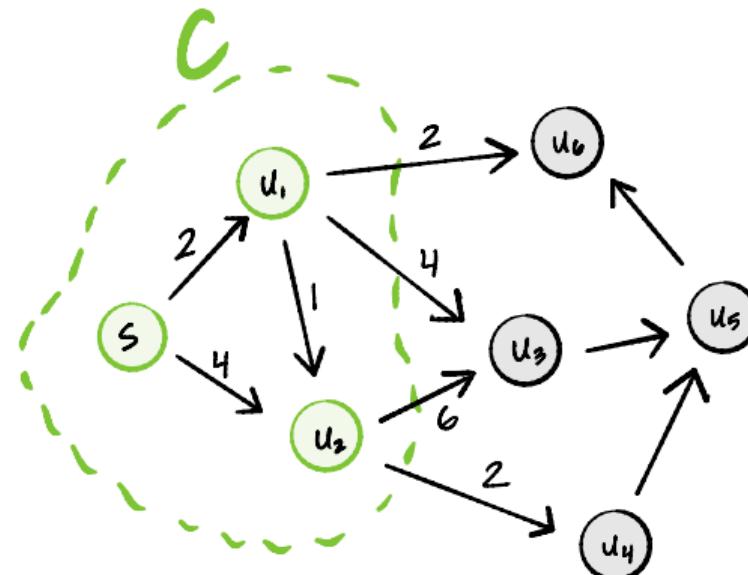
# Exit Paths

- True or False: this is an exit path from  $s$  through  $C$ .



# Path Decomposition

- ▶ Any path from  $s$  to a node  $u$  outside of  $C$  can be broken into two parts:  
(an exit path from  $s$ ) + (path from exit node to  $u$ )

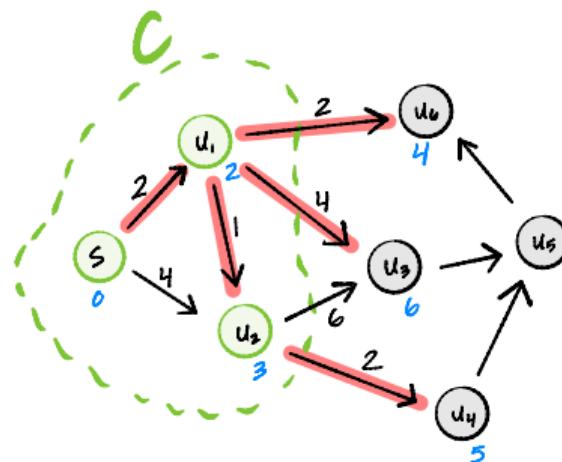


# Path Decomposition

- ▶ Consider any path from  $s$  to  $u \notin C$ .
- ▶ Suppose  $e$  is the path's exit node.
- ▶ We have:
  - (length of the path)
    - $= (\text{length of exit path to } e) + (\text{length of path from } e \text{ to } u)$
    - $\geq (\text{length of shortest exit path to } e) + (\text{length of path from } e \text{ to } u)$
- ▶ Since edge weights are positive, all path lengths  $\geq 0$ :
  - $\geq (\text{length of shortest exit path to } e) + 0$

# Shortest Exit Paths

- ▶ Example: What is the shortest exit path with exit node  $u_3$ ?



- ▶ If  $u$  is outside of  $C$ , then the length of the shortest exit path with exit node  $e$  is  $\text{est}[e]$ .

# Proof Idea

- ▶ Suppose  $u$  is a node outside of  $C$  for which  $\text{est}[u]$  is smallest.
- ▶ Consider any path from  $s$  to  $u$ , and let  $e$  be the path's exit node.
- ▶ We have:
$$\begin{aligned} & (\text{length of this path from } s \text{ to } u) \\ & \geq (\text{length of shortest exit path to } e) + 0 \\ & = \text{est}[e] \\ & \geq \text{est}[u] \end{aligned}$$
- ▶ That is, any path from  $s$  to  $u$  has length  $\geq \text{est}[u]$ .
- ▶ We've already found one with length  $\text{est}[u]$ ; this proves that it is the shortest.

# Outline of Dijkstra's Algorithm

```
def dijkstra(graph, weights, source):
    est = {node: float('inf') for node in graph.nodes}
    est[source] = 0
    pred = {node: None for node in graph.nodes}

    # empty set
    C = set()

    # while there are nodes still outside of C
        # find node u outside of C with smallest
        # estimated distance
    C.add(u)
    for v in graph.neighbors(u):
        update(u, v, weights, est, pred)

return est, pred
```

# Dijkstra's Algorithm: Naïve Implementation

```
1 def dijkstra(graph, weights, source):
2     est = {node: float('inf') for node in graph.nodes}
3     est[source] = 0
4     pred = {node: None for node in graph.nodes}
5
6     outside = set(graph.nodes)
7
8     while outside:
9         # find smallest with linear search
10        u = min(outside, key=est)
11        outside.remove(u)
12        for v in graph.neighbors(u):
13            update(u, v, weights, est, pred)
14
15    return est, pred
```

# Priority Queues

- ▶ A **priority queue** allows us to store (key, value) pairs, efficiently return key with lowest value.
- ▶ Suppose we have a priority queue class:
  - ▶ `PriorityQueue(priorities)` will create a priority queue from a dictionary whose values are priorities.
  - ▶ The `.extract_min()` method removes and returns key with smallest value.
  - ▶ The `.change_priority(key, value)` method changes key's value.

# Example

```
>>> pq = PriorityQueue({  
    'w': 5,  
    'x': 4,  
    'y': 1,  
    'z': 3  
)  
>>> pq.extract_min()  
'y'  
>>> pq.change_priority('w', 2)  
>>> pq.extract_min()
```

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# Dijkstra's Algorithm: Priority Queue

```
def dijkstra(graph, weights, source):
    est = {node: float('inf') for node in graph.nodes}
    est[source] = 0
    pred = {node: None for node in graph.nodes}

    priority_queue = PriorityQueue(est)
    while priority_queue:
        u = priority_queue.extract_min()
        for v in graph.neighbors(u):
            changed = update(u, v, weights, est, pred)
            if changed:
                priority_queue.change_priority(v, est[v])

    return est, pred
```

# Heaps

- ▶ A priority queue can be implemented using a **heap**.
- ▶ If a **binary min-heap** is used:
  - ▶ `PriorityQueue( est )` takes  $\Theta(V)$  time.
  - ▶ `.extract_min( )` takes  $O(\log V)$  time.
  - ▶ `.change_priority( )` takes  $O(\log V)$  time.

# Time Complexity Using Min Heap

```
def dijkstra(graph, weights, source):
    est = {node: float('inf') for node in graph.nodes}
    est[source] = 0
    pred = {node: None for node in graph.nodes}

    priority_queue = PriorityQueue(est)
    while priority_queue:
        u = priority_queue.extract_min()
        for v in graph.neighbors(u):
            changed = update(u, v, weights, est, pred)
            if changed:
                priority_queue.change_priority(v, est[v])

    return est, pred
```

- ▶ Time complexity: \_\_\_\_\_

