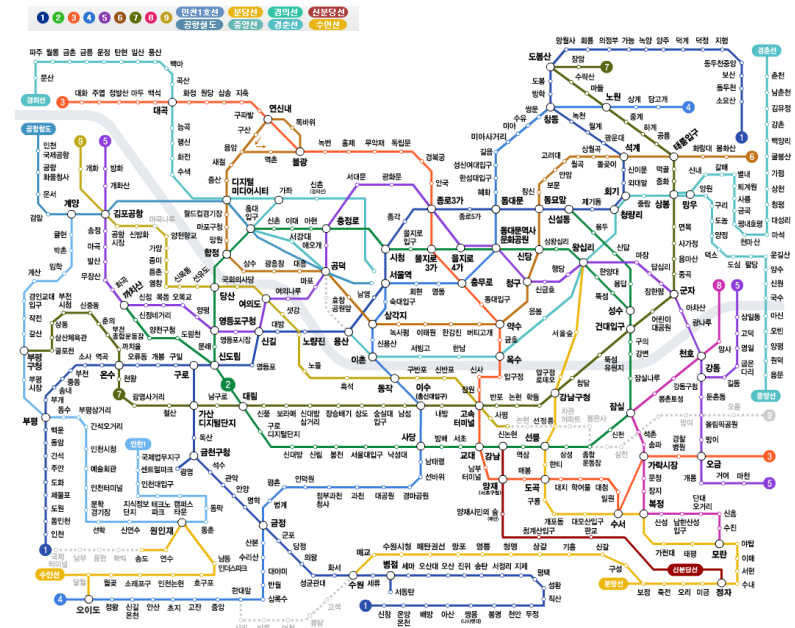


[CSED233-01] Data Structure

Shortest Path

Jaesik Park

POSTECH



Shortest Path Problems

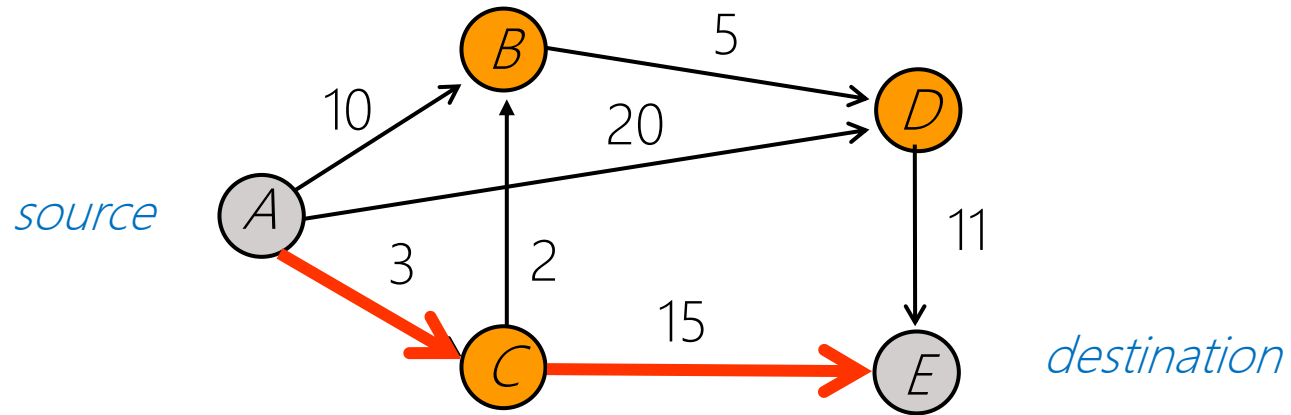
- **Weighted** digraph
 - Each edge (i,j) has a nonnegative cost $C[i,j]$
 - Also called weights or distances
- Path length/cost
 - Sum of weights of edges on path
 - source vertex \leftrightarrow destination vertex
- Shortest path problems (SPP)
 - Finding a path b/w two vertices such that the path length is minimized

Shortest Path Problems: Types

- **Single-pair** SPP
 - Single source, single destination
- **Single-source** SPP
 - Single source, all destinations
- **Single-destination** SPP
 - All sources, single destination
 - Can be reduced to the single-source SPP by reversing the edges in the graph
- **All-pairs** SPP
 - Every vertex is a source & destination

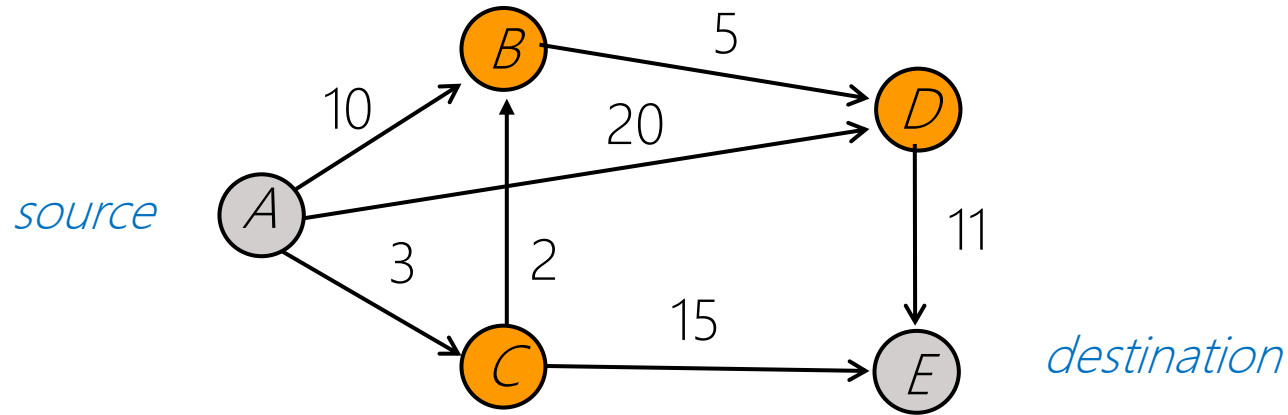
Single-Pair SPP: Single Source, Single Destination

- What is the shortest path from A to E ?



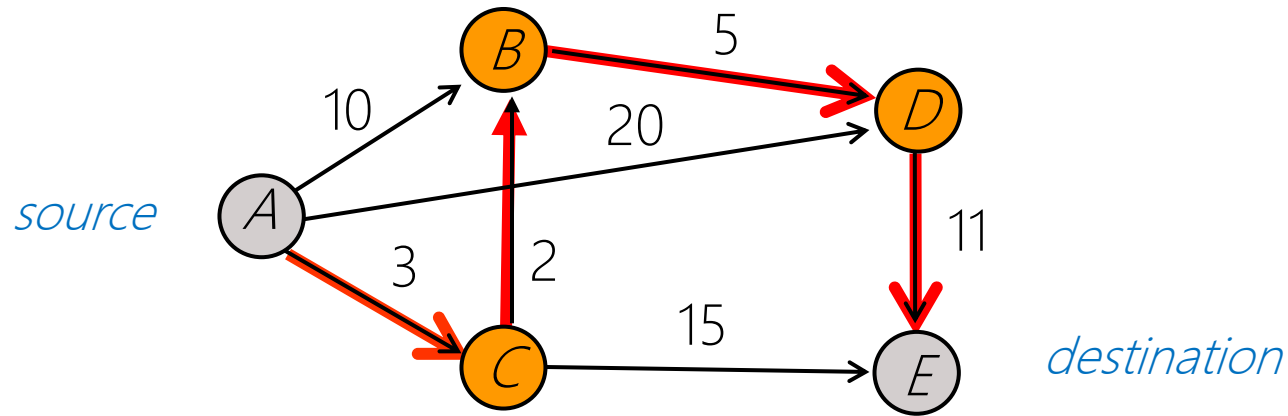
Shortest path length = 18

Single-Pair SPP: Single Source, Single Destination



- Greedy algorithm
 - Making the **locally-optimal** choice at each stage with the hope of finding a **global optimum**

Greedy Shortest A-to-E Path

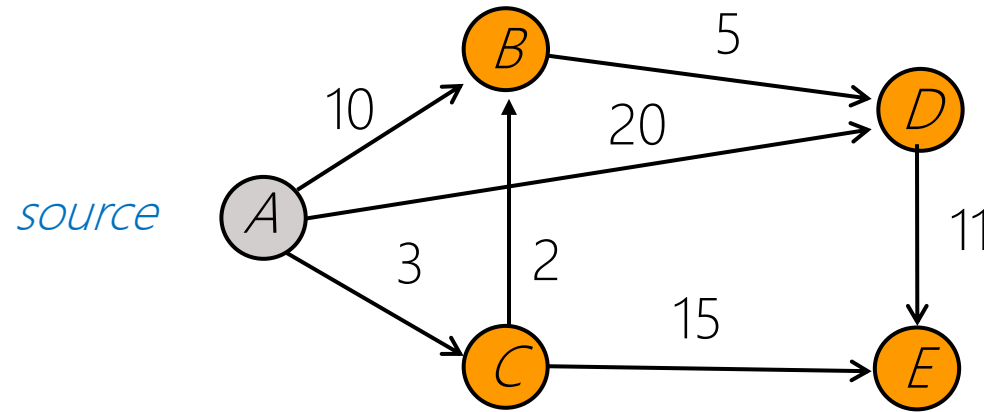


- Possible algorithm
 - Start with source vertex
 - Select a vertex using the cheapest edge subject to the constraint that a new vertex is reached
 - Continue until destination is reached
- Path length = 21 (not shortest path)
- Algorithm **doesn't work**

Single-Source SPP: Single Source, All Destinations

- Need to generate *n shortest paths*
 - From a given source to all destinations (*n* : # of vertices)
- Dijkstra's greedy method:
 - Generate the shortest paths *in stages*
 - Each stage generates a *shortest path* to a new destination
 - Greedy criterion at each stage
 - Select the *destination* (for the next shortest path) *in increasing order of its length*

Greedy: Single Source All Destinations



| Length | Shortest Path |
|--------|---------------|
| 0 | |
| 3 | |
| 5 | |
| 10 | |
| 18 | |

As a consequence,
each path (other than
first) is *one-edge
extension* of an already
generated shortest
path

Greedy: Single Source All Destinations

- $D[i]$ (distance from source to vertex i)
 - The length of the shortest one-edge extension of an already generated path, which ends at vertex i
 - Alternatively, $D[i]$ can be defined to be the length of the "current" shortest path to i (that passes only through vertices in S)
 - where S is a set of vertices whose shortest path (distance) from source is already known
- $P(i)$ (predecessor of i)
 - The vertex just before i on the shortest one-edge extension to i

Dijkstra's Alg. (Only to Compute Distance)

```
S := {1};  
for i := 2 to n do begin  
  D[i] := C[1, i]  
for i := 1 to n-1 do begin One-edge  
extension  
  choose a vertex  $w \in V-S$  whose  $D[.]$  value is least;  
  add  $w$  to  $S$ ;  
  Update of  $D[.]$   
  for each vertex  $v \in V-S$  do  
     $D[v] := \min(D[v], D[w] + C[w, v])$   
end;
```

Dijkstra's Alg. (To Record Actual Path)

```
S := {1};
```

```
for i := 2 to n do begin
```

```
    D[i] := C[1, i];
```

```
    if C[1, i]  $\neq \infty$  then P[i] := 1 end;
```

Initialize P[.]

```
for i := 1 to n-1 do begin
```

```
    choose a vertex  $w \in V-S$  whose D[.] value is least;
```

```
    add w to S;
```

```
    for each vertex  $v \in V-S$  do
```

```
        if (D[w] + C[w, v]) < D[v] then begin
```

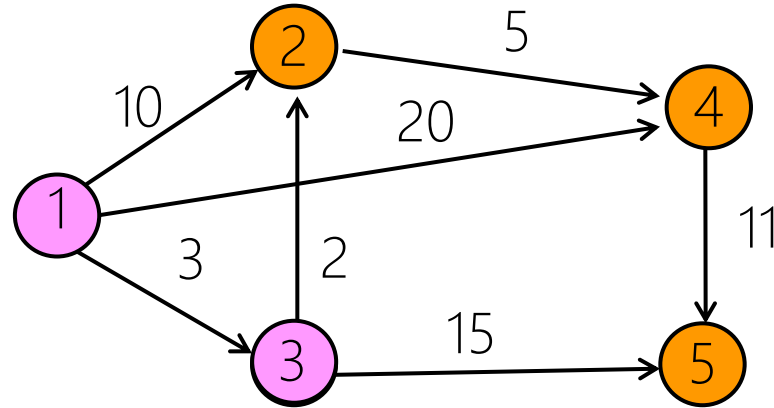
```
            D[v] := D[w] + C[w, v];
```

```
            P[v] := w end;
```

```
end;
```

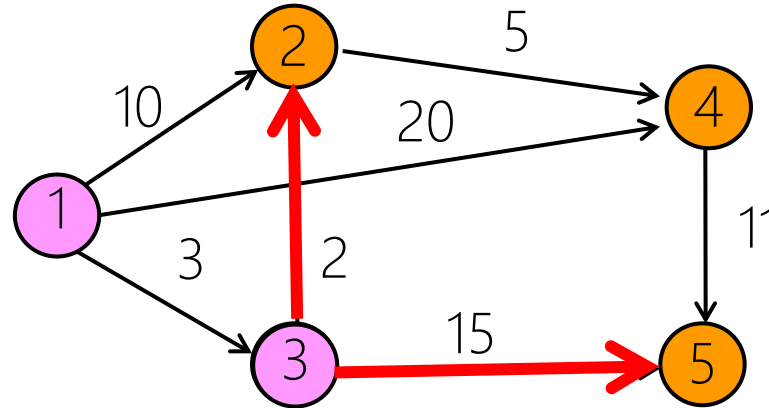
*Path
recording*

Progress of Dijkstra's Algorithm



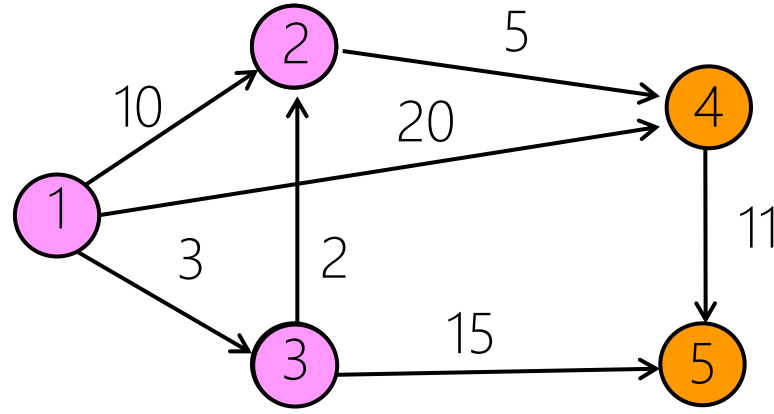
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|-----|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | | | | | |

Progress of Dijkstra's Algorithm



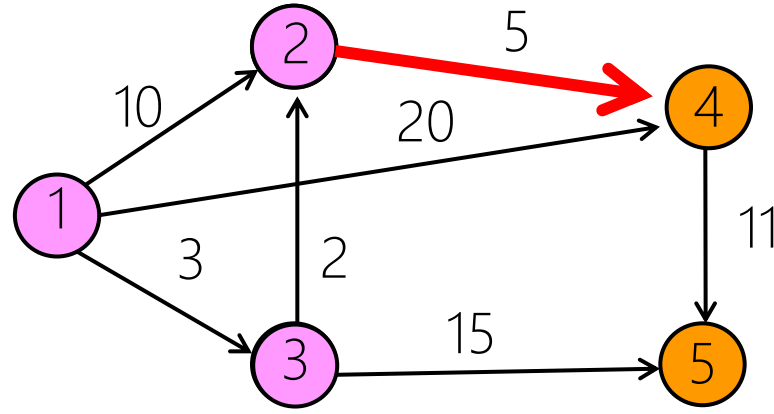
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|--------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |

Progress of Dijkstra's Algorithm



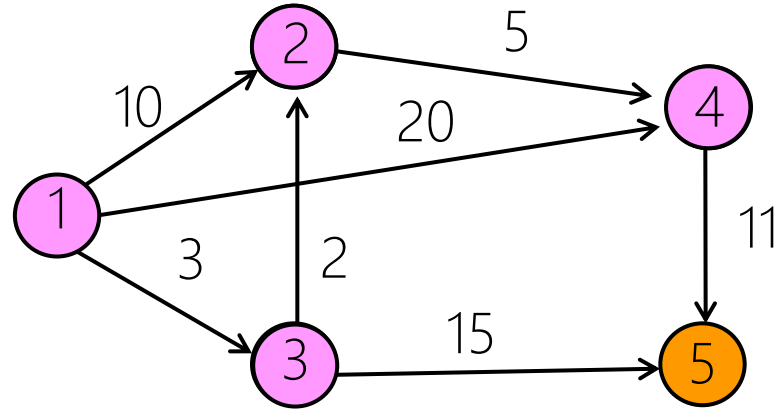
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|--------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | | | | | |

Progress of Dijkstra's Algorithm



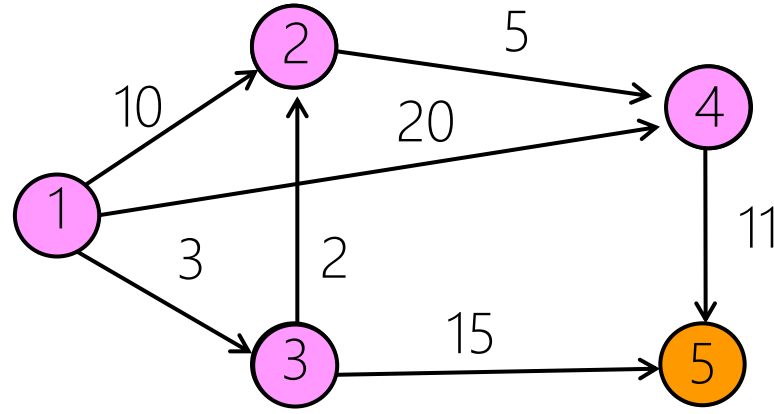
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|-----------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | {1, 3, 2} | 5/3 | 3/1 | 10/2 | 18/3 |

Progress of Dijkstra's Algorithm



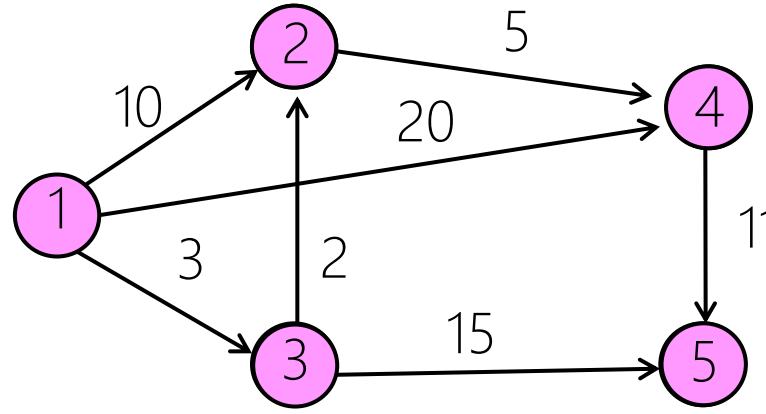
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|-----------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | {1, 3, 2} | 5/3 | 3/1 | 10/2 | 18/3 |
| 3 | | | | | |

Progress of Dijkstra's Algorithm



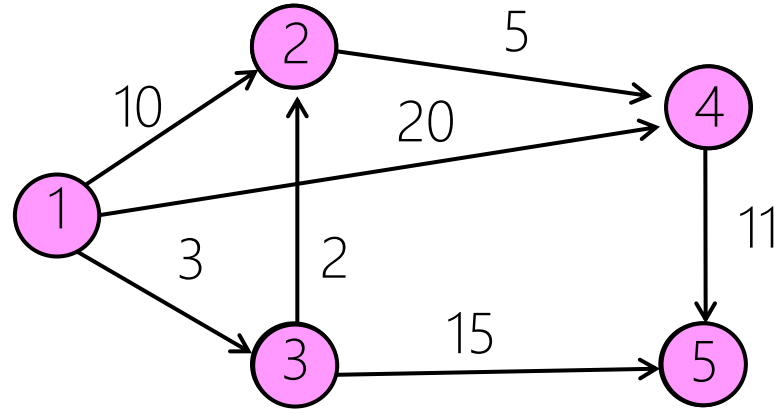
| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|--------------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | {1, 3, 2} | 5/3 | 3/1 | 10/2 | 18/3 |
| 3 | {1, 3, 2, 4} | 5/3 | 3/1 | 10/2 | 18/3 |

Progress of Dijkstra's Algorithm



| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|--------------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | {1, 3, 2} | 5/3 | 3/1 | 10/2 | 18/3 |
| 3 | {1, 3, 2, 4} | 5/3 | 3/1 | 10/2 | 18/3 |
| 4 | | | | | |

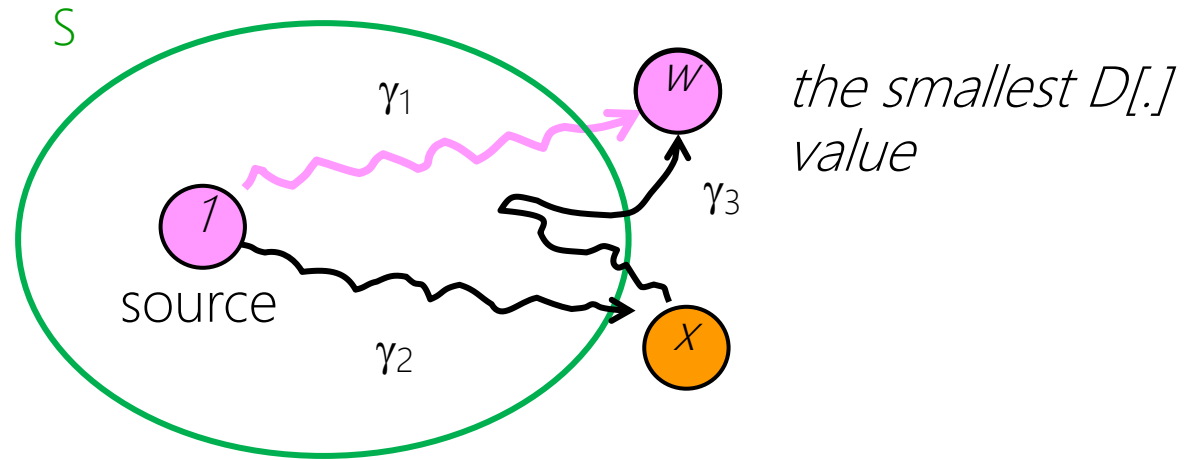
Progress of Dijkstra's Algorithm



| Stage | S | D/P[2] | D/P[3] | D/P[4] | D/P[5] |
|---------|-----------------|--------|--------|--------|--------------|
| Initial | {1} | 10/1 | 3/1 | 20/1 | ∞ / - |
| 1 | {1, 3} | 5/3 | 3/1 | 20/1 | 18/3 |
| 2 | {1, 3, 2} | 5/3 | 3/1 | 10/2 | 18/3 |
| 3 | {1, 3, 2, 4} | 5/3 | 3/1 | 10/2 | 18/3 |
| 4 | {1, 3, 2, 4, 5} | 5/3 | 3/1 | 10/2 | 18/3 |

Why Dijkstra's Algorithm Works (1)

- Locally-best choice turns out to be the best over all
 - Called **greedy choice property**



- If there were a shorter path, passing through a vertex x outside of S
 - that is, $|\gamma_2| + |\gamma_3| < |\gamma_1|$
 - Then $|\gamma_2| < |\gamma_1|$
 - thus the vertex x should have been selected before w
- **Contradiction**

Why Dijkstra's Algorithm Works (2)

- Update of $D[v]$ correctly keeps track of the shortest path to vertex v

- [See Aho83]



Complexity

- Depending on how to select the minimum D value?
- Option-1: **Scanning** the list of all vertices
 - $O(n \times n)$ to select next destination for all n vertices
 - $O(e)$ to update D & P values for all e edges with non-infinity cost $C[,]$ (using adjacency lists)

→ Total time: $O(n^2 + e) = O(n^2)$
- Option-2: **Using a min-heap** of D values
 - $O(n \log n)$ for n *DeleteMin* operations
 - $O(e \log n)$ for e *PriorityUpdate* operations

→ Total time: $O((n + e) \log n) = O(e \log n)$

 - Better for **sparse** graph: $O(n \log n)$

Further Questions

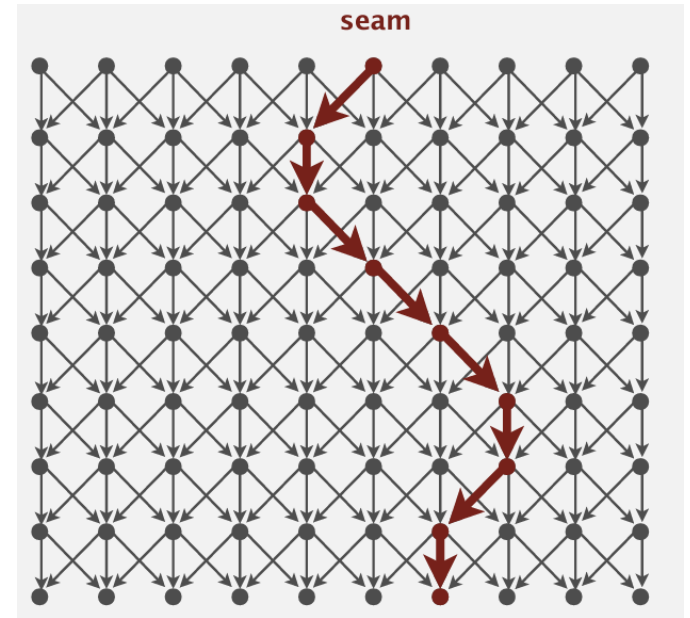
- Dijkstra's algorithm works with cyclic graphs?
 - Yes, why?
- What about negative edges?
 - No, why?
- Could we do better for acyclic graphs?
 - Use topological sort



Edsger W. Dijkstra
Turing award 1972

Application

- Content-aware resizing

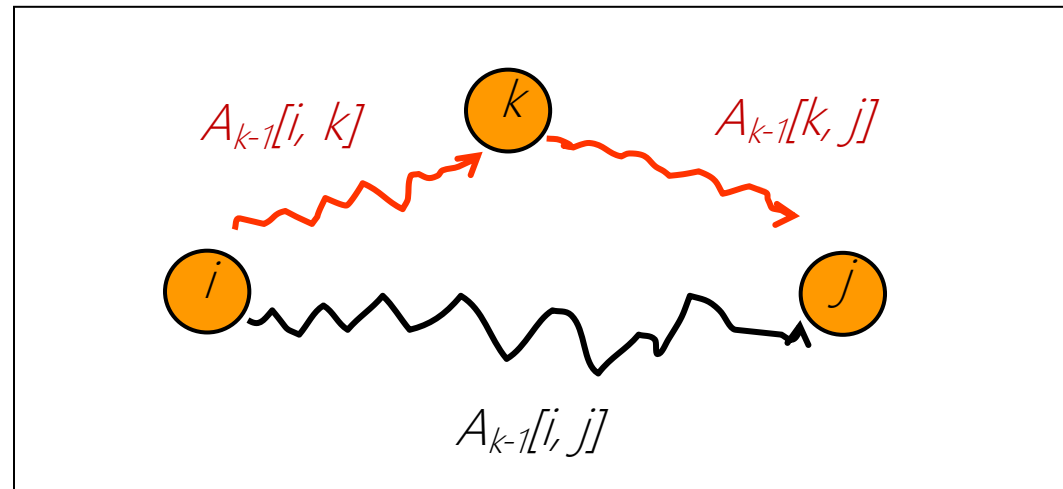


All-Pairs SPP

- One solution is to run **Dijkstra's** greedy algorithm n times
 - If G is **sparse** ($e = O(n)$), this is a good solution
 - Total time (of the min-heap version)
 - $O(n [e \log n]) = O(n^2 \log n)$
- Dynamic-programming solution by **Floyd**
 - Let $A_k[i,j]$ to be the shortest length of any path from vertex i to vertex j , whose intermediate vertices all have indices $\leq k$

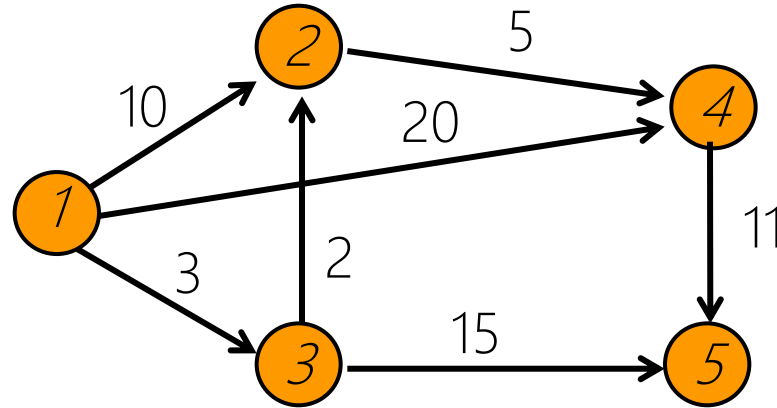
Floyd's Algorithm

- At the k -th iteration, we compute $A_k[.,.]$
 - $A_k[i, j] := \min\{ A_{k-1}[i, j], A_{k-1}[i, k] + A_{k-1}[k, j] \}$



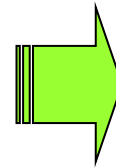
- Total time: $O(n^3)$

Example



| | 1 | 2 | 3 | 4 | 5 |
|---|---|----|---|----|----|
| 1 | 0 | 10 | 3 | 20 | - |
| 2 | - | 0 | - | 5 | - |
| 3 | - | 2 | 0 | - | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_0[i, j]$



| | 1 | 2 | 3 | 4 | 5 |
|---|---|---|---|---|---|
| 1 | | | | | |
| 2 | | | | | |
| 3 | | | | | |
| 4 | | | | | |
| 5 | | | | | |

$A_5[i, j]$

Space Reduction

- $A_k[i, j] := \min\{ A_{k-1}[i, j], A_{k-1}[i, k] + A_{k-1}[k, j] \}$
 - 3-dimensional space
- In all cases, $A_k[i, j]$ can overwrite $A_{k-1}[i, j]$
 - When i equals k , $A_k[k, j] = A_{k-1}[k, j]$
 - When j equals k , $A_k[i, k] = A_{k-1}[i, k]$
 - When neither i nor j equals k ,
 $A_{k-1}[i, j]$ is used only in the computation of $A_k[i, j]$

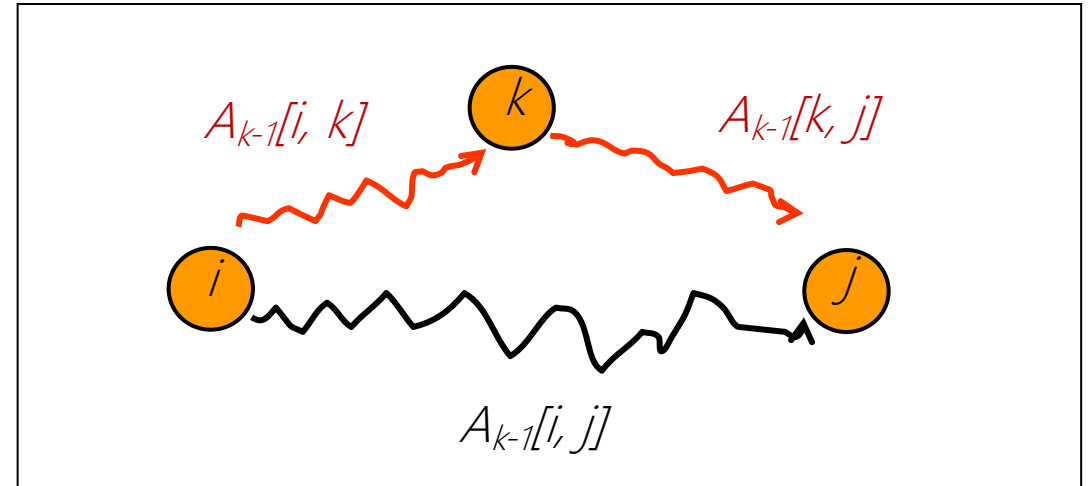
Building The Shortest Paths



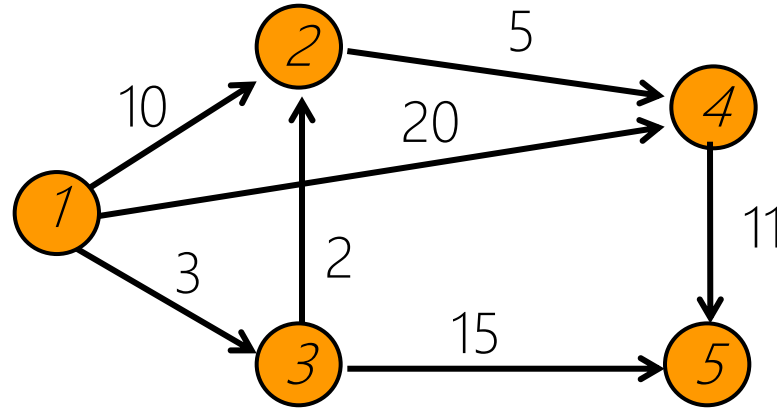
[See Aho83]

Building The Shortest Paths

- If the path exists between two nodes then $\text{Next}[u][v] = v$
else we set $\text{Next}[u][v] = -1$
- $\text{if}(A[i][j] > A[i][k] + A[k][j])$
 {
 $A[i][j] = A[i][k] + A[k][j];$
 $\text{Next}[i][j] = \text{Next}[i][k];$
 }
- $\text{path} = [u]$
 while $u \neq v$:
 $u = \text{Next}[u][v]$
 $\text{path.append}(u)$



Example



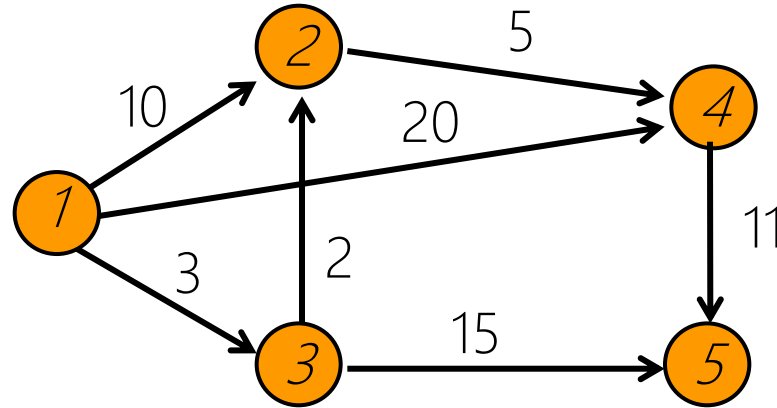
| | 1 | 2 | 3 | 4 | 5 |
|---|---|----|---|----|----|
| 1 | 0 | 10 | 3 | 20 | - |
| 2 | - | 0 | - | 5 | - |
| 3 | - | 2 | 0 | - | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_o[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 2 | 3 | 4 | -1 |
| 2 | -1 | -1 | -1 | 4 | -1 |
| 3 | -1 | 2 | -1 | -1 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_o[i, j]$

Example



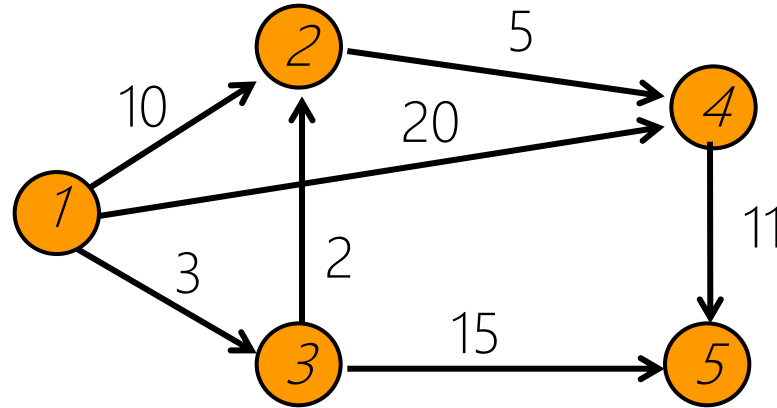
| | 1 | 2 | 3 | 4 | 5 |
|---|---|----|---|----|----|
| 1 | 0 | 10 | 3 | 20 | - |
| 2 | - | 0 | - | 5 | - |
| 3 | - | 2 | 0 | - | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_1[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 2 | 3 | 4 | -1 |
| 2 | -1 | -1 | -1 | 4 | -1 |
| 3 | -1 | 2 | -1 | -1 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_1[i, j]$

Example



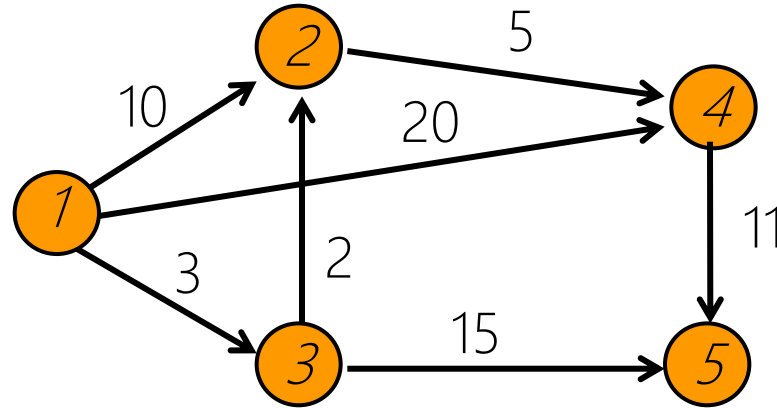
| | 1 | 2 | 3 | 4 | 5 |
|---|---|----|---|----|----|
| 1 | 0 | 10 | 3 | 15 | - |
| 2 | - | 0 | - | 5 | - |
| 3 | - | 2 | 0 | 7 | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_2[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 2 | 3 | 2 | -1 |
| 2 | -1 | -1 | -1 | 4 | -1 |
| 3 | -1 | 2 | -1 | 2 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_2[i, j]$

Example



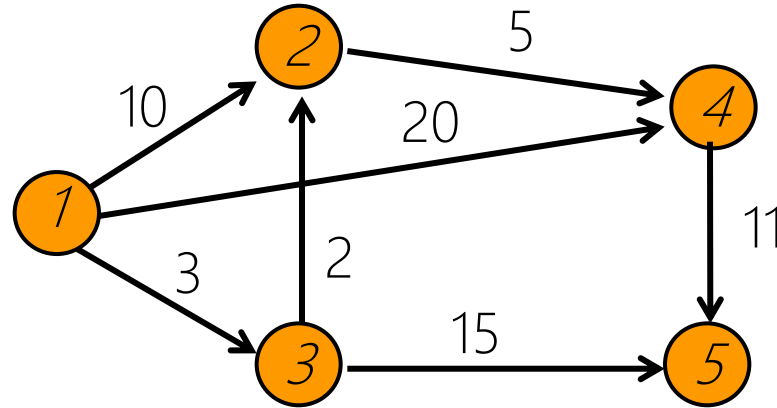
| | 1 | 2 | 3 | 4 | 5 |
|---|---|---|---|----|----|
| 1 | 0 | 5 | 3 | 10 | 18 |
| 2 | - | 0 | - | 5 | - |
| 3 | - | 2 | 0 | 7 | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_3[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 3 | 3 | 3 | 3 |
| 2 | -1 | -1 | -1 | 4 | -1 |
| 3 | -1 | 2 | -1 | 2 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_3[i, j]$

Example



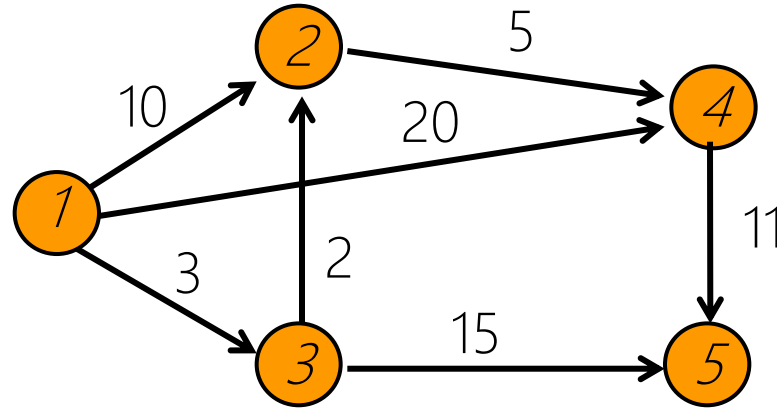
| | 1 | 2 | 3 | 4 | 5 |
|---|---|---|---|----|----|
| 1 | 0 | 5 | 3 | 10 | 18 |
| 2 | - | 0 | - | 5 | 16 |
| 3 | - | 2 | 0 | 7 | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

$A_4[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 3 | 3 | 3 | 3 |
| 2 | -1 | -1 | -1 | 4 | 4 |
| 3 | -1 | 2 | -1 | 2 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_4[i, j]$

Example



| | 1 | 2 | 3 | 4 | 5 |
|---|---|---|---|----|----|
| 1 | 0 | 5 | 3 | 10 | 18 |
| 2 | - | 0 | - | 5 | 16 |
| 3 | - | 2 | 0 | 7 | 15 |
| 4 | - | - | - | 0 | 11 |
| 5 | - | - | - | - | 0 |

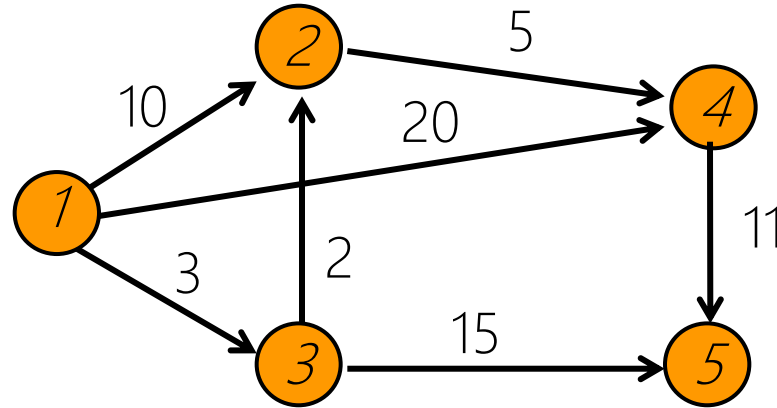
$A_5[i, j]$

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 3 | 3 | 3 | 3 |
| 2 | -1 | -1 | -1 | 4 | 4 |
| 3 | -1 | 2 | -1 | 2 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$Next_5[i, j]$

Example

```
path = [u]
while u != v:
    u = Next[u][v]
    path.append(u)
```



Shortest path from 1 to 4?
Path = [1]

$u = \text{Next}[1,4] = 3$
Path = [1,3]

$u = \text{Next}[3,4] = 2$
Path = [1,3,2]

$u = \text{Next}[2,4] = 4$
Path = [1,3,2,4]

| | 1 | 2 | 3 | 4 | 5 |
|---|----|----|----|----|----|
| 1 | -1 | 3 | 3 | 3 | 3 |
| 2 | -1 | -1 | -1 | 4 | 4 |
| 3 | -1 | 2 | -1 | 2 | 5 |
| 4 | -1 | -1 | -1 | -1 | 5 |
| 5 | -1 | -1 | -1 | -1 | -1 |

$A_5[i, j]$

$\text{Next}_5[i, j]$

Comparison of Floyd's with Dijkstra's

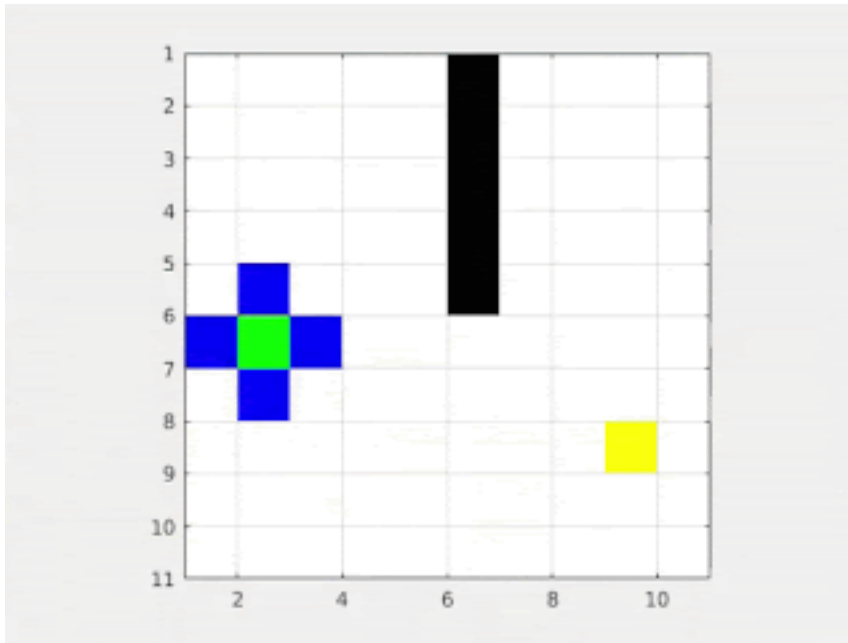
- Floyd's is better for a **dense** graph
 - $O(n^3)$ vs. $O(n^3 \log n)$
- Floyd's algorithm
 - Works even when the graph has a **negative** edge cost (if there are no negative-length cycles)

Other Shortest Path Algorithms (1)

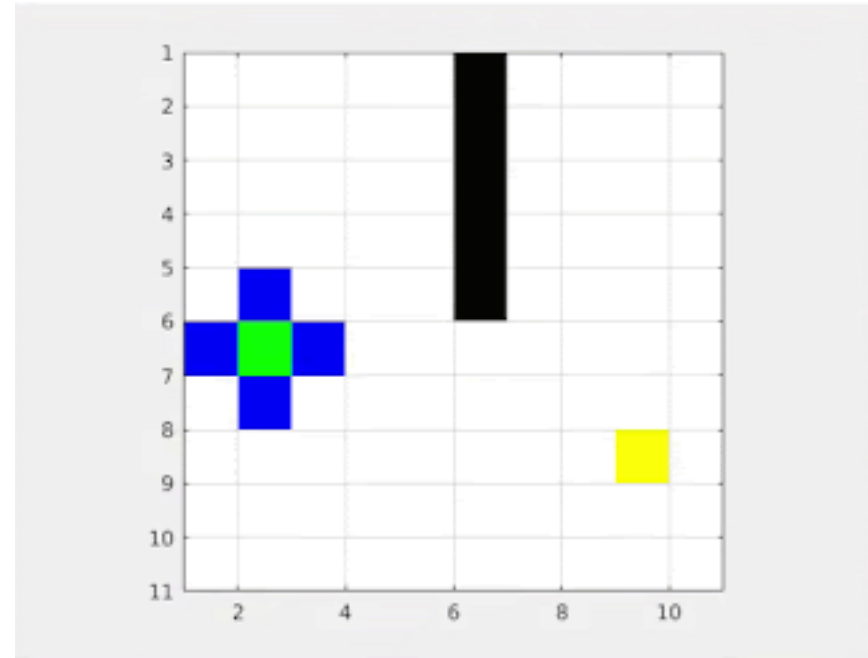
- **Bellman-Ford** algorithm
 - Solves **single-source** SPP (allowing **negative** edge costs)
 - Time complexity: $O(ne)$
 - Slower than Dijkstra's
- **Johnson's** algorithm
 - Solves **all-pairs** SPP (allowing **negative** edge costs, but no negative-length cycles)
 - Time complexity: $O(n^2 \log n + ne)$
 - May be faster than Floyd's on sparse graphs

Other Shortest Path Algorithms (2)

- A^* search algorithm
 - Solves single-pair SPP
 - Uses heuristics to try to speed up the search



Dijkstra



A^*

References

- Further reading list and references
 - <https://www.geeksforgeeks.org/dijkstras-shortest-path-algorithm-greedy-algo-7/>
 - <https://www.geeksforgeeks.org/a-search-algorithm/>
- Slide credit
 - Jaesik Park
 - Seung-Hwan Baek
 - Jong-Hyeok Lee