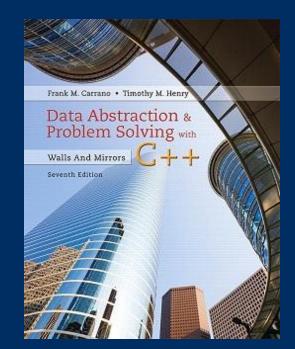
Graphs – Shortest Path



CS 302 - Data Structures

M. Abdullah Canbaz



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Reminders

- Assignment 7 is available
 - Due Wednesday, May 7th at 2pm
 - -TA
 - Athanasia Katsila,
 Email: akatsila [at] nevada {dot} unr {dot} edu,
 Office Hours: Thursdays, 10:30 am 12:30 pm at SEM 211
- Assignment 8 is available
 - Due Wednesday, May 16th at 2pm
 - -TA
 - Shehryar Khattak,
 Email: shehryar [at] nevada {dot} unr {dot} edu,
 Office Hours: Friday, 11:00 am 1:00 pm at ARF 116



Graph Algorithms

Depth-first search

 Visit all the nodes in a branch to its deepest point before moving up

Breadth-first search

 Visit all the nodes on one level before going to the next level

Single-source shortest-path

 Determines the shortest path from a designated starting node to every other node in the graph



Single Source Shortest Path





Shortest-path problem

- There might be multiple paths from a source vertex to a destination vertex
- Shortest path: the path whose total weight (i.e., sum of edge weights) is minimum

Austin→Dallas→Denver→Atlanta→Washington: 2980 miles



Variants of Shortest Path

- Single-pair shortest path
 - Find a shortest path from u to v
 - for given vertices u and v

- Single-source shortest paths
 - G = (V, E) \Rightarrow find a shortest path from a given source vertex **s** to each vertex **v** ∈ **V**



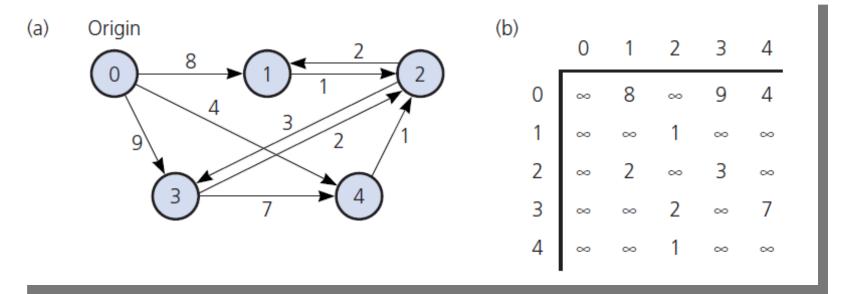
Variants of Shortest Paths (cont'd)

- Single-destination shortest paths
 - Find a shortest path to a given destination vertex t
 from each vertex v
 - Reversing the direction of each edge → single-source

- All-pairs shortest paths
 - Find a shortest path from u to v for every pair of vertices u and v



- The shortest path between two vertices in a weighted graph
 - Has the smallest edge-weight sum

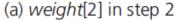


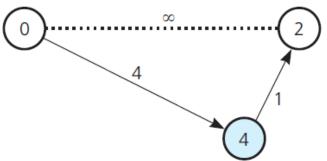
(a) A weighted directed graph and (b) its adjacency matrix



| | | | weight | | | | |
|------|---|---------------|--------|-----|-----|-----|------|
| Step | V | vertexSet | [0] | [1] | [2] | [3] | _[4] |
| 1 | _ | 0 | 0 | 8 | 00 | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | 7 | 5 | 8 | 4 |
| 4 | 1 | 0, 4, 2, 1 | 0 | 7 | 5 | 8 | 4 |
| 5 | 3 | 0, 4, 2, 1, 3 | 0 | 7 | 5 | 8 | 4 |
| | | | | | | | |

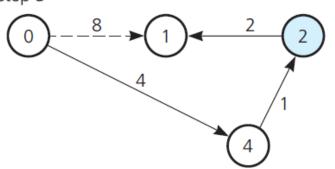
 A trace of the shortest-path algorithm applied to the graph





Step 2. The path 0–4–2 is shorter than 0–2

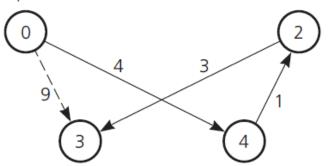
(b) weight[1] in step 3



Step 3. The path 0-4-2-1 is shorter than 0-1

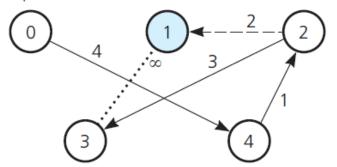
- Checking weight[u] by examining the graph:
 - (a) weight[2] in step 2; (b) weight[1] in step 3;

(c) weight[3] in step 3



Step 3 continued. The path 0–4–2–3 is shorter than 0–3

(d) weight[3] in step 4



Step 4. The path 0-4-2-3 is shorter than 0-4-2-1-3

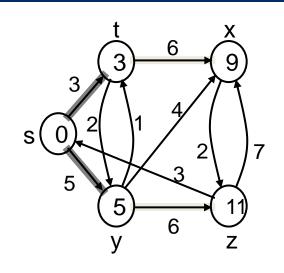
Checking weight[u] by examining the graph:
 (c) weight[3] in step 3; (d) weight[3] in step 4

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Notation

• Weight of path $p = \langle v_0, v_1, \dots, v_k \rangle$

$$w(p) = \sum_{i=1}^{k} w(v_{i-1}, v_i)$$

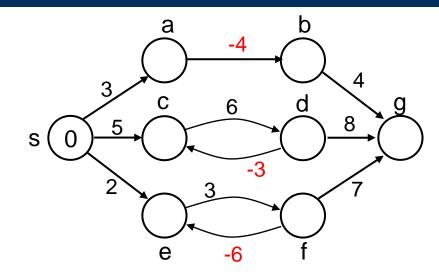


Shortest-path weight from s to V:

$$\delta(v) = \begin{cases} \min w(p) : s \xrightarrow{p} v \text{ if there exists a path from s to } v \\ \infty & \text{otherwise} \end{cases}$$

Negative Weights and Negative Cycles

 Negative-weight edges may form negative-weight cycles.



- If negative cycles are reachable from the source, the shortest path is not well defined.
 - i.e., keep going around the cycle, and get $w(s, v) = -\infty$ for all v on the cycle



Could shortest path solutions contain cycles?

- Negative-weight cycles
 - Shortest path is not well defined
- Positive-weight cycles:
 - By removing the cycle, we can get a shorter path
- Zero-weight cycles
 - No reason to use them; can remove them to obtain a path with same weight



Shortest-path algorithms

- Solving the shortest path problem in a brute-force manner requires enumerating all possible paths.
 - There are O(V!) paths between a pair of vertices in an acyclic graph containing V nodes.

- We will discuss two algorithms
 - Dijkstra's algorithm
 - Bellman-Ford's algorithm

M Shortest-path algorithms

- Dijkstra's and Bellman-Ford's algorithms are "greedy" algorithms!
 - Find a "globally" optimal solution by making "locally" optimum decisions.
- Dijkstra's algorithm
 - Does not handle negative weights.
- Bellman-Ford's algorithm
 - Handles negative weights but not negative cycles reachable from the source.



Shortest-path algorithms (cont'd)

• Both Dijkstra's and Bellman-Ford's algorithms are iterative:

 Start with a shortest path estimate for every vertex: d[v]

 Estimates are updated iteratively until convergence:

$$d[v] \rightarrow \delta(v)$$

Shortest-path algorithms (cont'd)

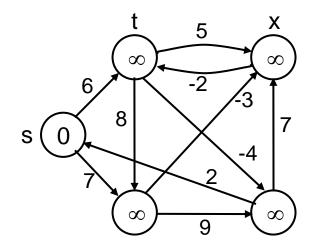
Two common steps:

(1) Initialization

(2) Relaxation (i.e., update step)

M Initialization Step

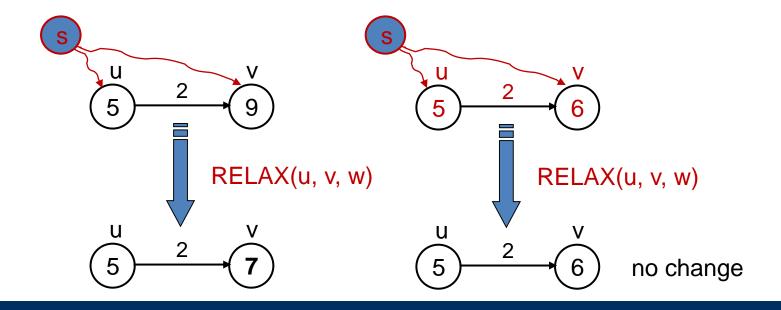
- Set d[s]=0
 - i.e., source vertex
- Set $d[v] = \infty$ for $v \neq s$
 - i.e., large value



Relaxation Step

Relaxing an edge (u, v) implies testing whether we can improve the shortest path to v found so far by going through u:

```
If d[v] > d[u] + w(u, v)
we can improve the shortest path to v
 \Rightarrow d[v]=d[u]+w(u,v)
```





Bellman-Ford Algorithm

Can handle negative weights

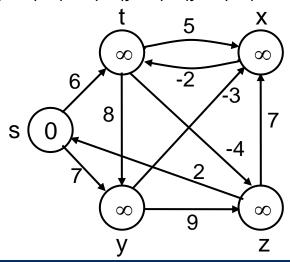
Detects negative cycles reachable from the source

 Returns FALSE if negative-weight cycles are reachable from the source s ⇒ no solution

| Bellman-Ford Algorithm (cont'd)

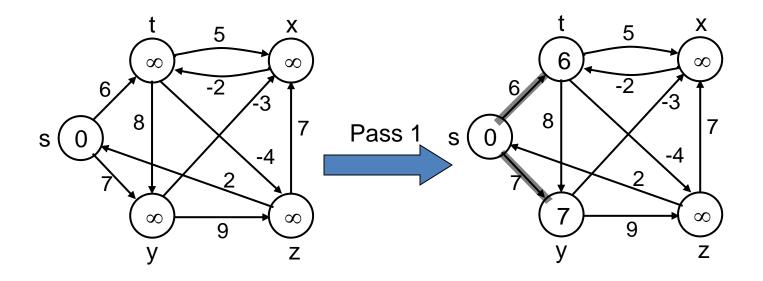
- Each edge is relaxed |V-1| times by making |V-1| passes over the whole edge set
 - to make sure that each edge is relaxed exactly |V-1| times
- it puts the edges in an unordered list and goes over the list |V-1| times

(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)





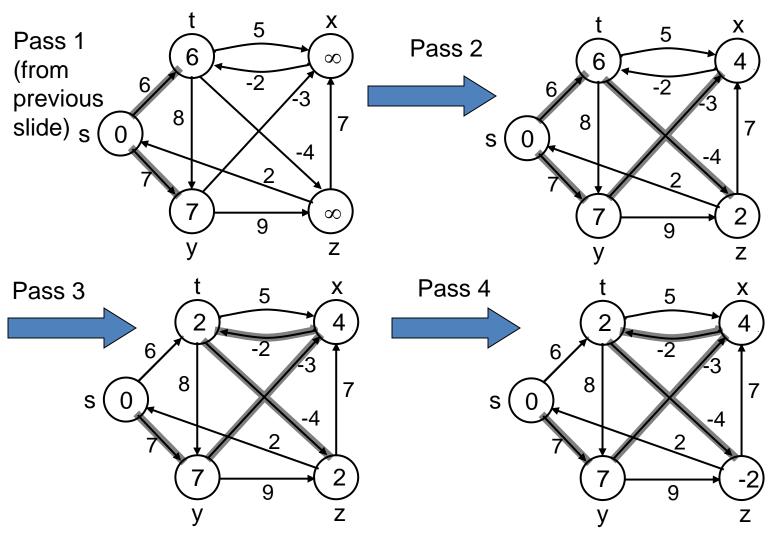
Example



E: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)

Example

(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)





Detecting Negative Cycles: needs an extra iteration

for each edge $(u, v) \in E$ do if d[v] > d[u] + w(u, v)then return FALSE

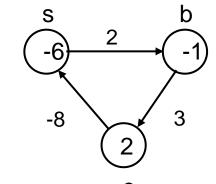
3

return TRUE

1st pass

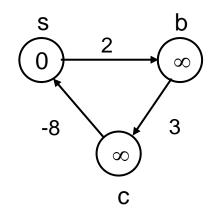
2

-8



2nd pass

(s,b) (b,c) (c,s)



Consider edge (s, b):

$$d[b] = -1$$

 $d[s] + w(s, b) = -4$

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BELLMAN-FORD Algorithm

- INITIALIZE-SINGLE-SOURCE(V, s) ← O(V)
 for i ← 1 to |V| 1
 for each edge (u, v) ∈ E ← O(E)
- 4. RELAX(u, v, w)
- 5. **for** each edge $(u, v) \in E$ $\longleftarrow O(E)$
- 6. **if** d[v] > d[u] + w(u, v)
- 7. **return** FALSE
- 8. **return** TRUE

Time: O(V+VE+E)=O(VE)

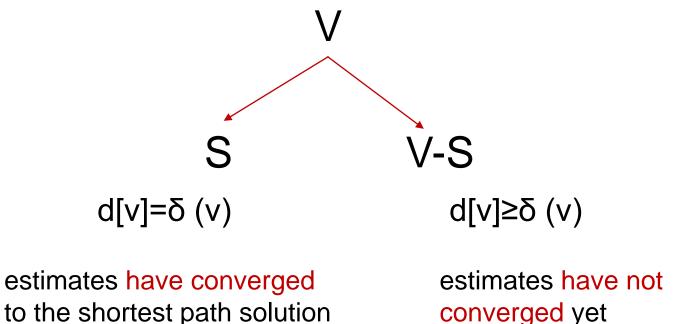
Dijkstra's Algorithm

- Cannot handle negative-weights!
 - w(u, v) > 0, \forall (u, v) \in E

Each edge is relaxed only once!

Dijkstra's Algorithm (cont'd)

At each iteration, it maintains two sets of vertices:



Initially, S is empty

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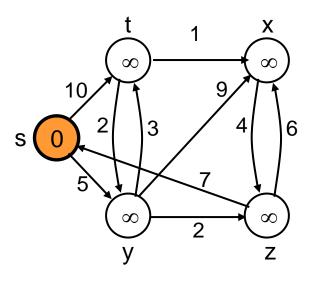
Dijkstra's Algorithm (cont.)

- Vertices in V–S reside in a min-priority queue Q
 - Priority of u determined by d[u]
 - The "highest" priority vertex will be the one having the smallest d[u] value.

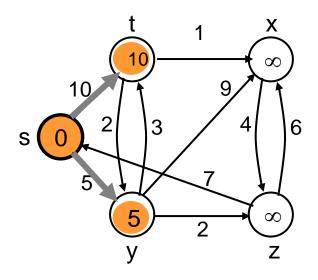
Steps

- 1) Extract a vertex u from Q
- 2) Insert u to S
- 3) Relax all edges leaving u
- Update Q

Dijkstra (G, w, s)



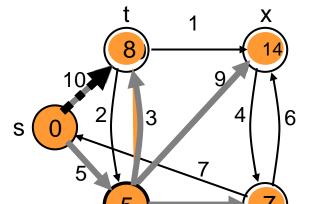




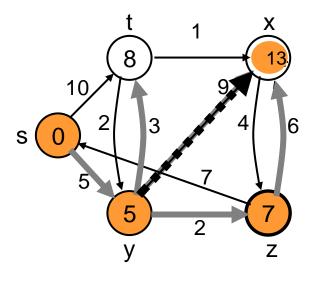


Example (cont.)

$$S=Q=$$

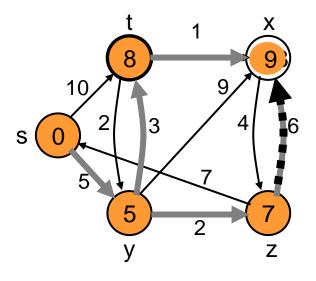


$$S= Q=$$

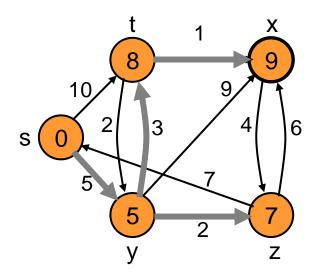


Example (cont.)

$$S=Q=$$



$$S=Q=<>$$



Note: use back-pointers to recover the shortest path solutions!

Dijkstra (G, w, s)

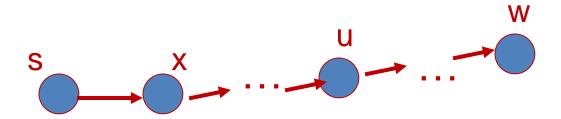
```
INITIALIZE-SINGLE-SOURCE(V, s)
                                                             \leftarrow O(V)
S \leftarrow \emptyset
                             build priority heap
                        \leftarrow O(V logV)
Q \leftarrow V[G]
while Q \neq \emptyset
                         \leftarrow O(V) times
      u \leftarrow EXTRACT-MIN(Q)
                                               \leftarrow O(logV)
       S \leftarrow S \cup \{u\}
                                                        \leftarrow O(E_{vi})
       for each vertex v \in Adi[u]
                                                                                O(E_{vi} log V)
            RELAX(u, v, w)
                                                           ← O(logV)
            Update Q (DECREASE KEY)
```

Overall: $O(V+2V\log V+(E_{v1}+E_{v2}+...)\log V) = O(V\log V+E\log V) = O(E\log V)$



Improving Dijkstra's efficiency

Suppose the shortest path from s to w is the following:



- If u is the i-th vertex in this path, it can be shown that $d[u] \rightarrow \delta$ (u) at the i-th iteration:
 - move u from V-S to S
 - d[u] never changes again

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Add a flag for efficiency!

```
INITIALIZE-SINGLE-SOURCE(V, s)
S \leftarrow \emptyset
Q \leftarrow V[G]
while Q \neq \emptyset
    u \leftarrow EXTRACT-MIN(Q)
       S \leftarrow S \cup \{u\}; \leftarrow mark u
       for each vertex v \in Adj[u]
             If v not marked
              RELAX(u, v, w)
              Update Q (DECREASE KEY)
```



Dijkstra vs Bellman-Ford

Bellman-Ford

Dijkstra

O(VE)
$$V^2$$
 if G is sparse: E=O(V) V^3 if G is dense: E=O(V^2)

O(ElogV) VlogV if G is sparse: E=O(V) $V^2logV \text{ if G is dense: } E=O(V^2)$



Revisiting BFS

- BFS can be used to solve the shortest path problem when the graph is <u>weightless</u> or when all the weights are equal.
 - Path with lowest number of edges
 - i.e., connections
- Need to "mark" vertices before Enqueue!
 - i.e., do not allow duplicates



Circuits

Circuit

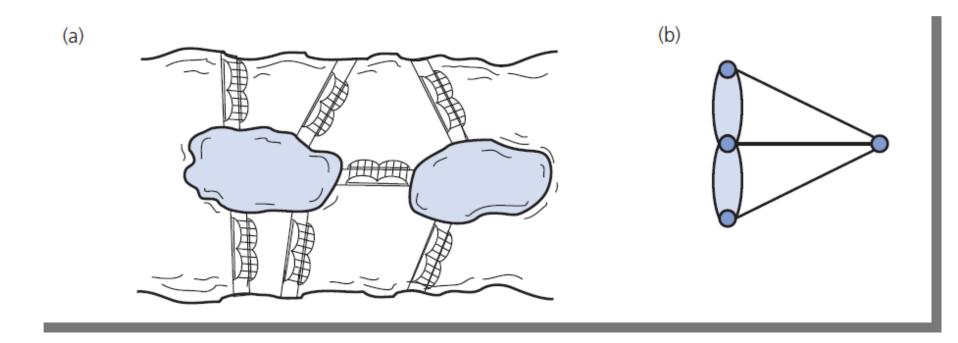
- Another name for type of cycle common in statement of certain types of problems
- Typical circuits either visit every vertex once or every edge once

Euler Circuit

- Begins at vertex v
- Passes through every edge exactly once
- Terminates at v



Circuits



(a) Euler's bridge problem and
(b) its multigraph representation



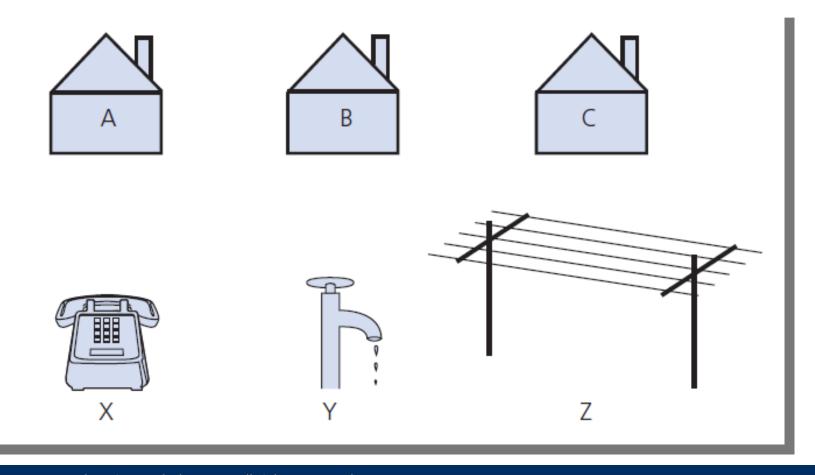
Some Difficult Problems

- Hamilton circuit
 - Begins at vertex v
 - Passes through every vertex exactly once
 - Terminates at v
- Variation is "traveling salesperson problem"
 - Visit every city on his route exactly once
 - Edge (road) has associated cost (mileage)
 - Goal is determine least expensive circuit



Some Difficult Problems

The three utilities problem



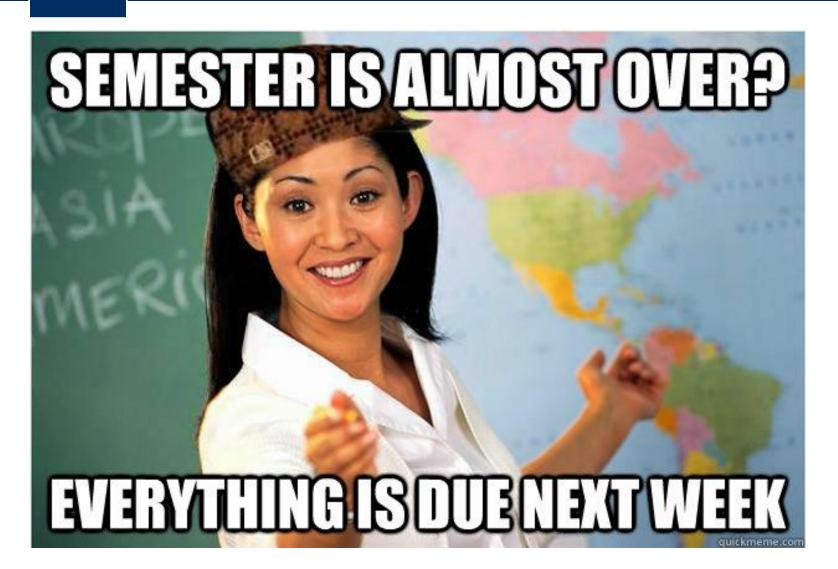


Remember





also





and





finally



Its due until 11:59 PM on Wed, May 9, 2018.



