

Lecture 8

Single-Source Shortest Paths Problems Continued

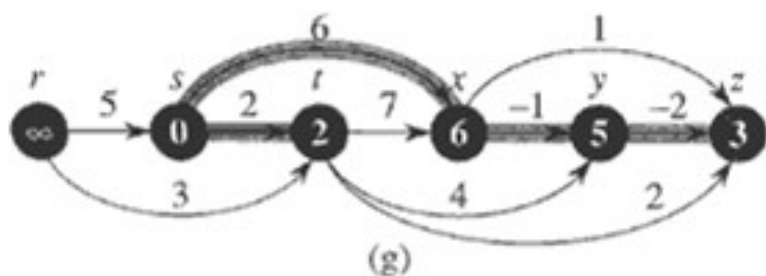
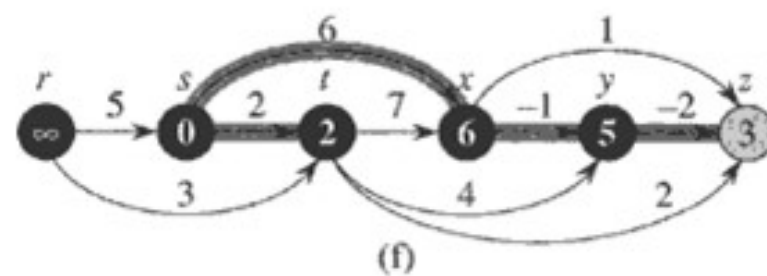
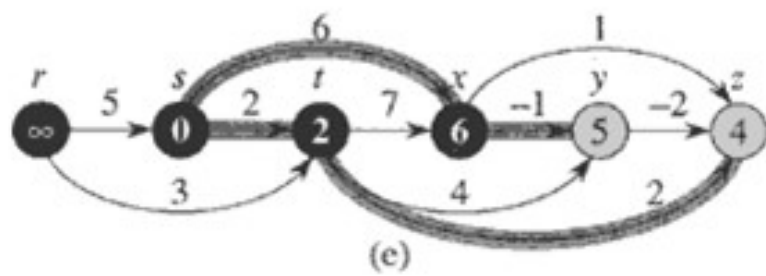
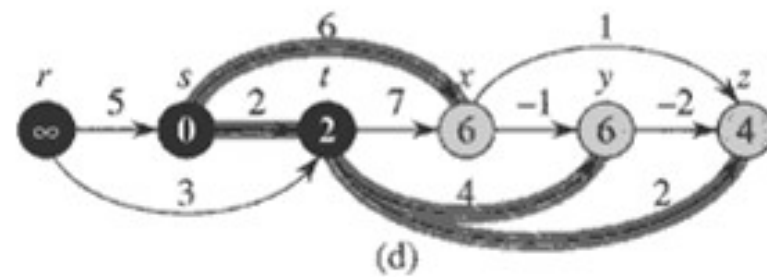
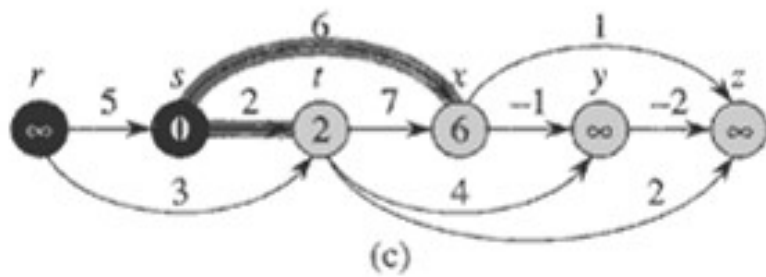
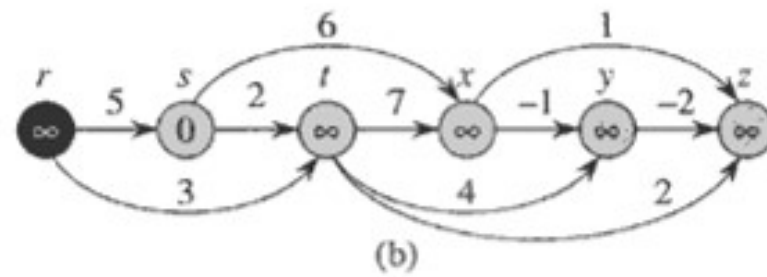
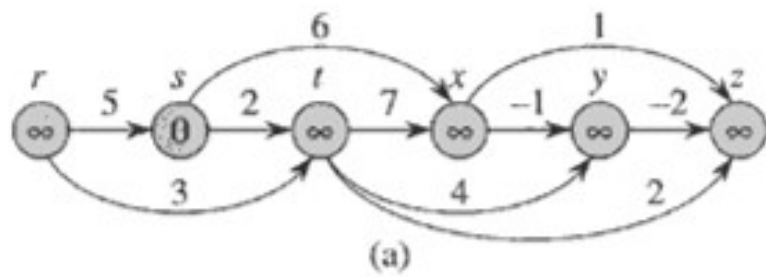
1. Single-Source Shortest Paths in Directed Acyclic Graphs
2. Dijkstra's Algorithm
3. Difference constraints

Single-Source Shortest Paths in Directed Acyclic Graphs

DAG-SHORTEST-PATHS(G, w, s)

- 1 topologically sort the vertices of G
- 2 INITIALIZE-SINGLE-SOURCE(G, s)
- 3 **for** each vertex u , taken in topologically sorted order
- 4 **do for** each vertex $v \in Adj[u]$
- 5 **do** RELAX(u, v, w)

Time Complexity: $O(V + E)$



Correctness of the Algorithm

- Theorem 24.5

- If a weighted, directed graph $G = (V, E)$ has source vertex s and no cycles, then at the termination of the DAG-SHORTEST-PATHS procedure, $d[v] = \delta(s, v)$ for all vertices $v \in V$, and the predecessor subgraph G_π is a shortest-paths tree rooted at s .

- Proof.

- We first show that $d[v] = \delta(s, v)$ for all vertices $v \in V$ at the termination.
 - If v is not reachable from s , then $d[v] = \delta(s, v) = \infty$ by the no-path property.
-

Proof Continued

- Now, Suppose that there is a shortest path $P = \langle v_0, v_1, \dots, v_k \rangle$, where $v_0 = s$ and $v_k = v$. Since we process the vertices in topological order, the edges on P are relaxed in the order $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$. The path relaxation property implies that $d[v_i] = \delta(s, v_i)$ for all i .
 - By the predecessor-subgraph property, G_π is a shortest-paths tree.
-

Application: Determine Critical Path

- Pert Chart:

- edges represent jobs and weights represent the times required to perform particular jobs.
 - If edge (u,v) enters v and edge (v,w) leaves v , then job (u,v) must be performed prior to job (v,w) .
 - A path through this dag represents a sequence of jobs that must be performed in a particular order.
 - A critical path is a longest path through the dag.
-

Two method

- Negating the edge weight.
- Replace “ ∞ ” with “ $-\infty$ ” and “ $>$ ” with “ $<$ ”.

Dijkstra's Algorithm

- Problem solved
- Correctness
- Time complexity
- Similarities to BFS and PRIM

Problem Solved by Dijkstra's

- Single-source shortest-paths problem
 - Edge weight ≥ 0
 - Input: A graph $G=(V, E)$ and a source s , and a nonnegative function $w: E \rightarrow \mathbb{R}^+$
 - Output: For each vertex v , shortest path weight $\delta(s, v)$, and a shortest path if exists.
-

The Dijkstra Algorithm

DIJKSTRA(G, w, s)

1. **Initialize-Single-Source**(G, s)

2. $S \leftarrow \emptyset$

3. $Q \leftarrow V[G]$

4. **while** $Q \neq \emptyset$

5. **do** $u \leftarrow \text{EXTRACT-MIN}(Q)$

6. $S \leftarrow S \cup \{u\}$

7. **for each** $v \in \text{Adj}[u]$

8. **do** **RELAX**(u, v, w)

INITIALIZE-SINGLE-SOURCE(G, s)

1 **for each** vertex $v \in V[G]$

2 **do** $d[v] \leftarrow \infty$

3 $\pi[v] \leftarrow \text{NIL}$

4 $d[s] \leftarrow 0$

RELAX(u, v, w)

1 **if** $d[v] > d[u] + w(u, v)$

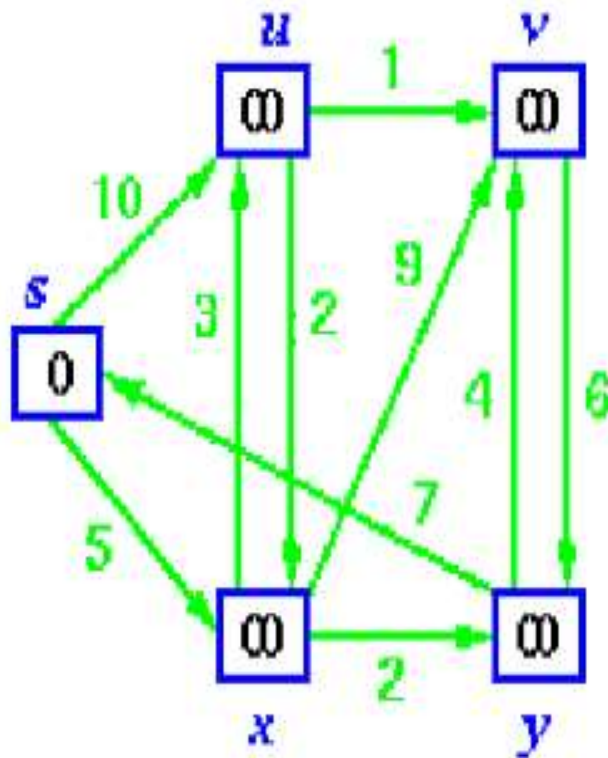
2 **then** $d[v] \leftarrow d[u] + w(u, v)$

3 $\pi[v] \leftarrow u$

Dijkstra's Algorithm - example

- Initial Graph

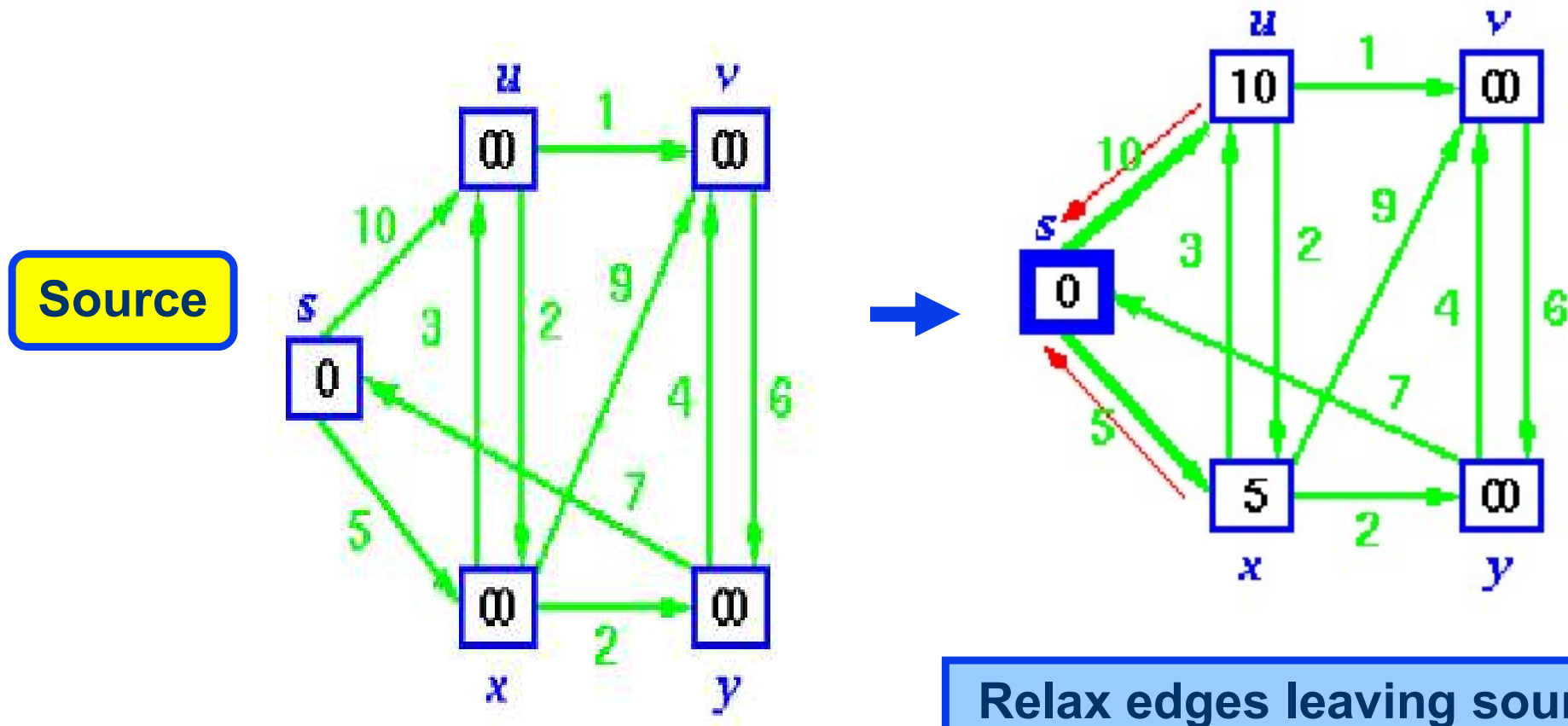
Source
Mark 0



Distance to all
nodes marked ∞

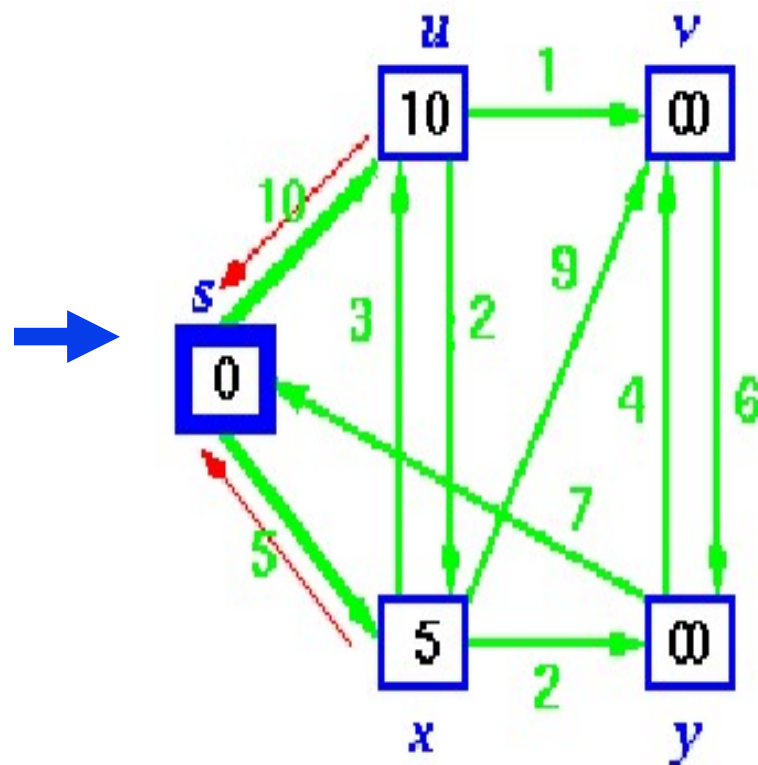
Dijkstra's Algorithm - Operation

- Initial Graph



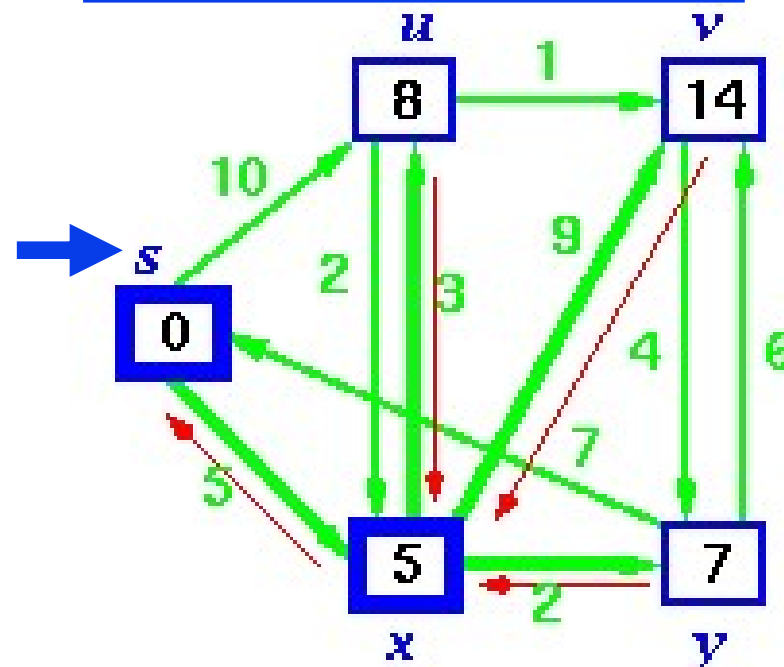
$$S = \emptyset$$

Dijkstra's Algorithm - Operation



Red arrows show predecessors

Relax u and a shorter path via x exists

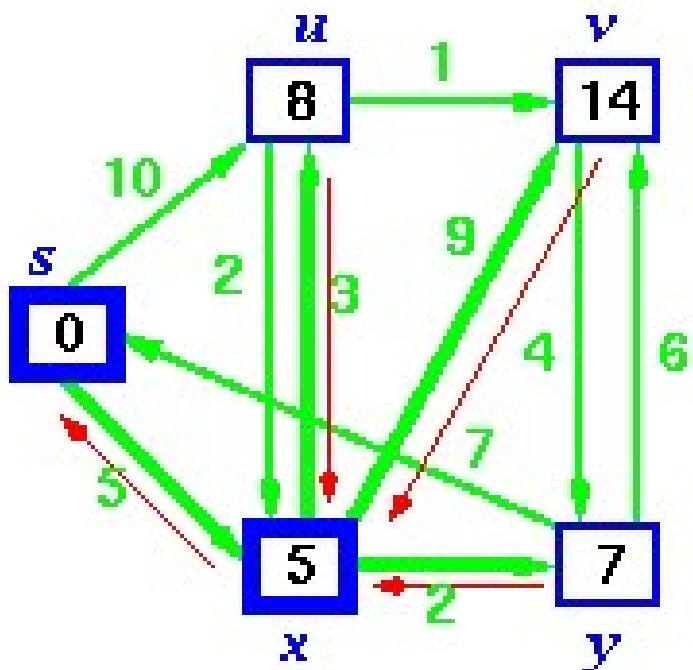


Sort vertices and choose closest

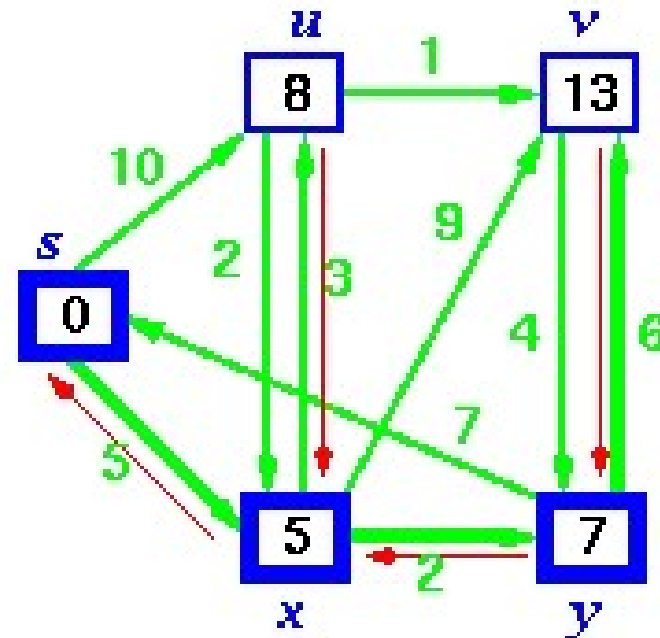
Relax y

Dijkstra's Algorithm - Operation

Relax v and a shorter path via y exists



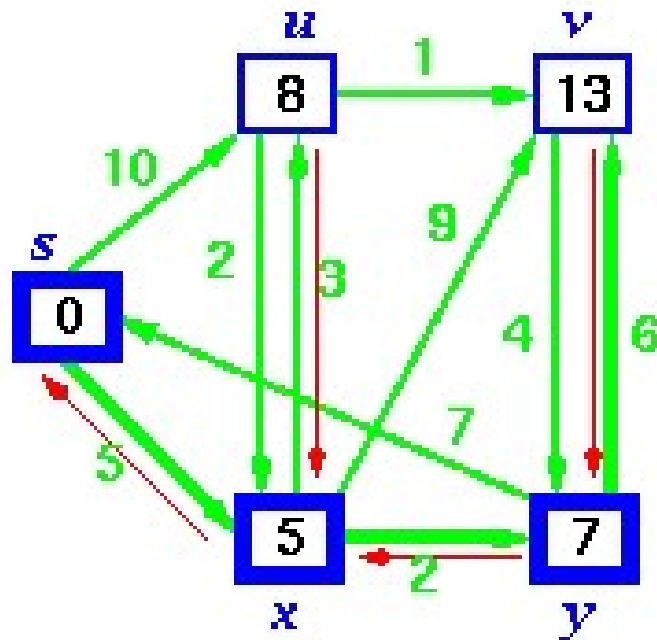
S is now $\{s, x\}$



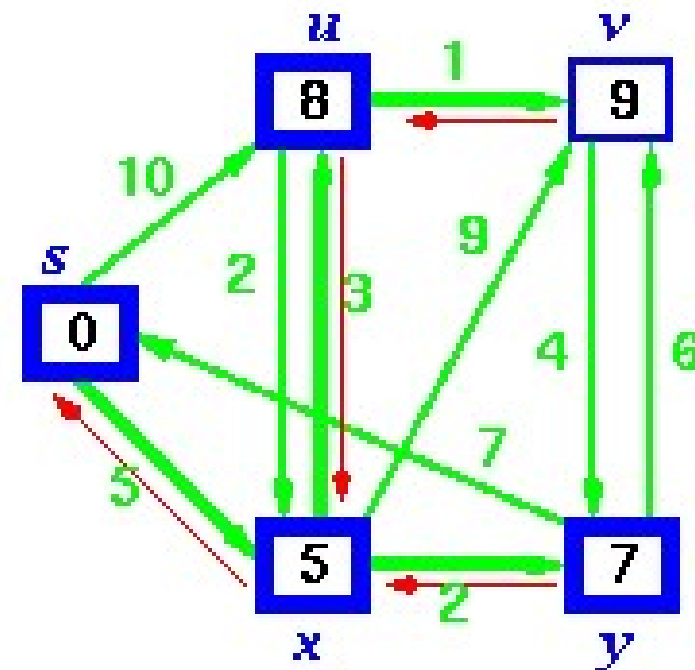
Sort vertices and choose closest

Dijkstra's Algorithm - Operation

Update $d[v]$

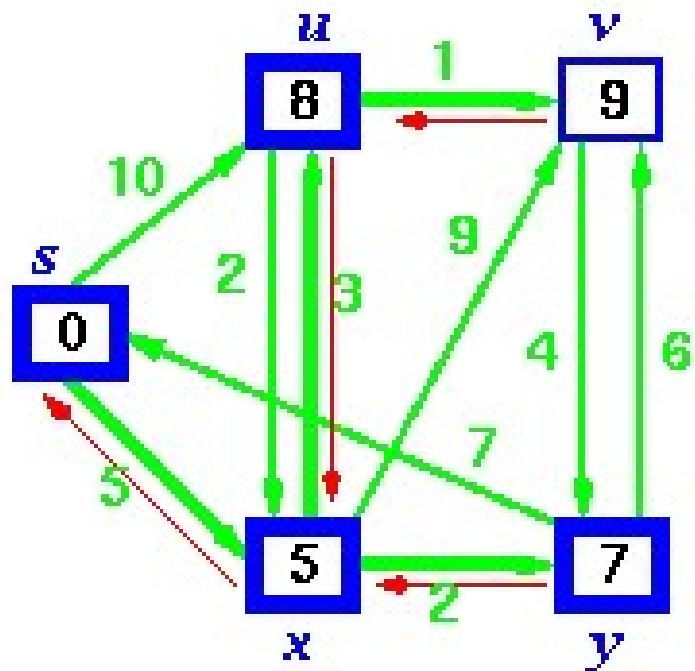


S is now $\{s, x, y\}$

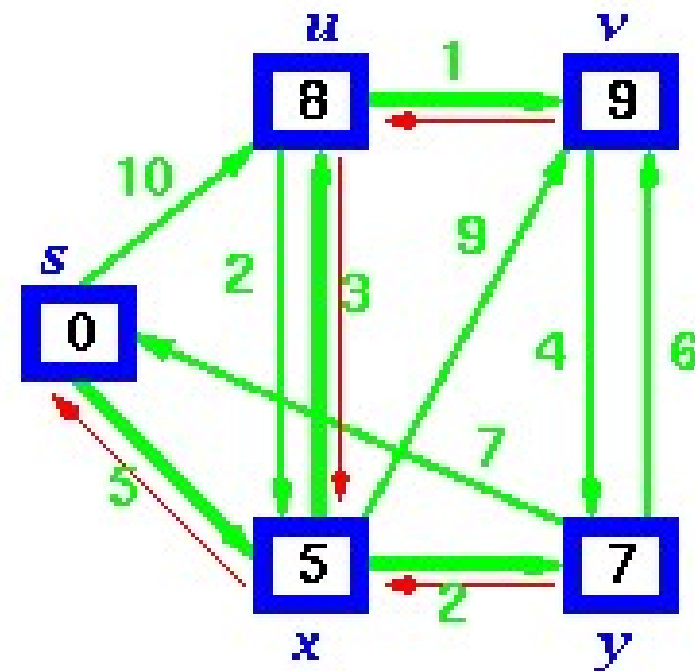


Sort vertices and choose closest, u

Dijkstra's Algorithm - Operation

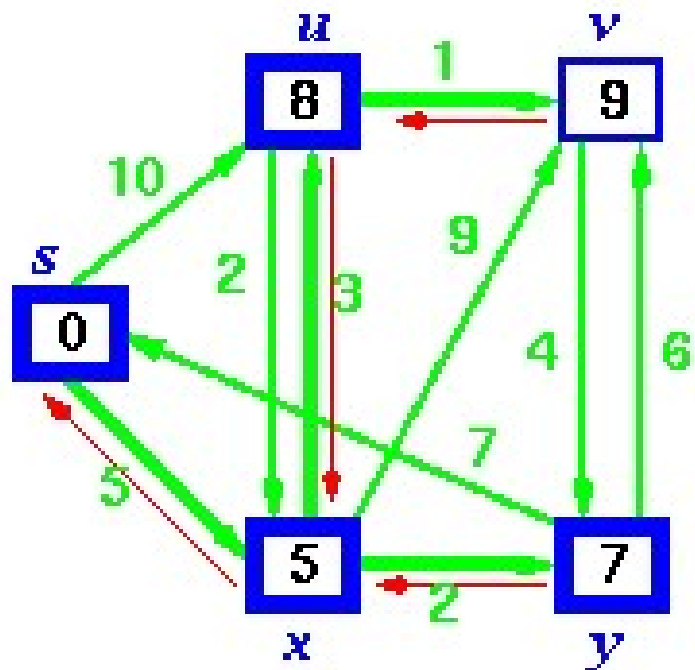


S is now $\{s, x, y, u\}$

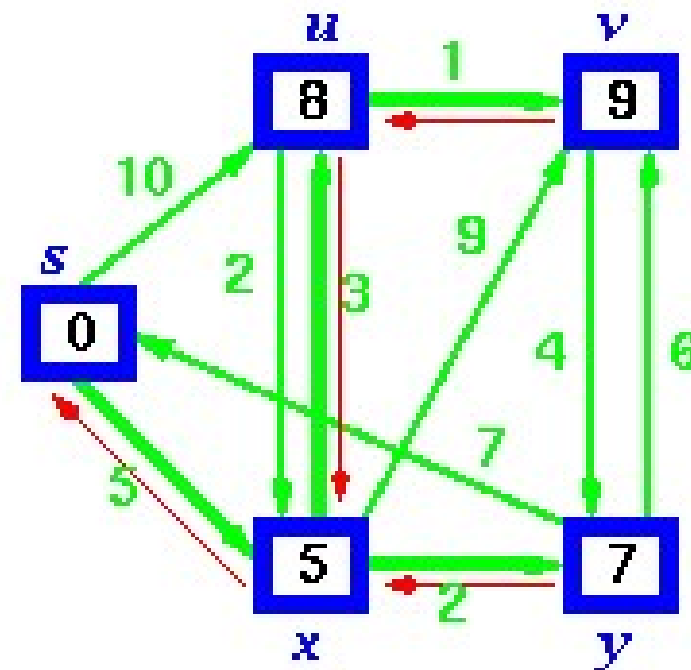


Finally add v

Dijkstra's Algorithm - Operation



S is now $\{s, x, y, u\}$



Pre-decessors show
shortest paths sub-graph

Correctness of Dijkstra's

- Theorem 24.6

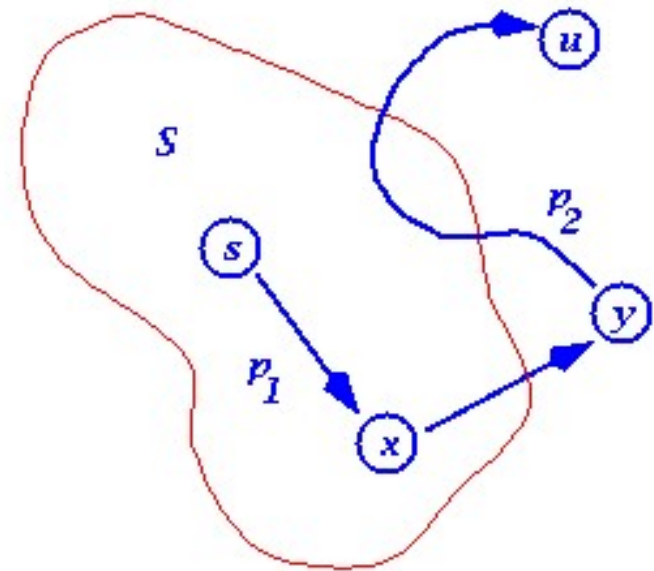
- Dijkstra's algorithm, run on a weighted, directed graph $G=(V, E)$, with non-negative weight function w and source s , terminates with $d[u] = \delta(s, u)$ for all vertices $u \in V$.

- Proof (by contradiction)

- Since $S=V$ in the end, we only need to prove that for each vertex v added to S , there holds $d[v] = \delta(s, v)$ when v is added to S .
- Suppose that **u is the first** vertex for which $d[u] \neq \delta(s, u)$ when it was added to S
- Note
 - u is not s because $d[s] = 0 = \delta(s, s)$
 - There must be a path $s \rightarrow \dots \rightarrow u$, since otherwise $d[u] = \delta(s, u) = \infty$.
 - Since there's a path, there must be a shortest path (note there is no negative cycle).

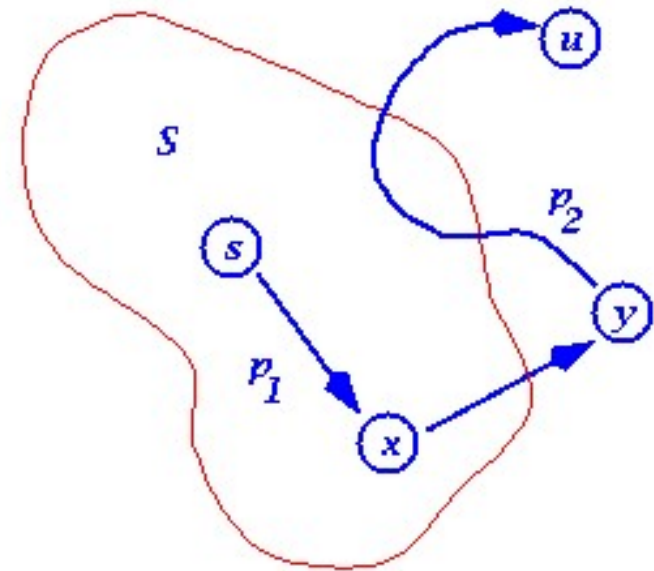
Dijkstra's Algorithm - Proof

- Let $s \rightsquigarrow x \rightarrow y \rightsquigarrow u$ be a shortest path from s to u , where at the moment u is chosen to S , x is in S and y is the first outside S
- When x was added to S , $d[x] = \delta(s, x)$
- Edge $x \rightarrow y$ was relaxed at that time, so at time u is chosen, $d[y] = \delta(s, y)$



Dijkstra's Algorithm - Proof

- So, $d[y] = \delta(s, y) \leq \delta(s, u) \leq d[u]$
- But, when we chose u , both u and y are in Q , so $d[u] \leq d[y]$ (otherwise we would have chosen y)
- Thus the inequalities must be equalities
 $\therefore d[y] = \delta(s, y) = \delta(s, u) = d[u]$
- And our hypothesis ($d[u] \neq \delta(s, u)$) is contradicted!



- 1. Why the red inequality holds? How about negative edges exist?
- 2. When u is added to S , a shortest path $s \rightarrow u$ in S is found.

Dijkstra's Algorithm - Time Complexity

- Like that of PRIM's
 - If using arrays
 - $O(V^2)$
 - Using special data structure
 - $O(V \lg V + E)$ or less
-

Compared to BFS

DIJKSTRA(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s)
2. $S \leftarrow \emptyset$
3. $Q \leftarrow V[G]$
4. while $Q \neq \emptyset$
do $u \leftarrow \text{EXTRACT-MIN}(Q)$
 $S \leftarrow S \cup \{u\}$
for each $v \in \text{Adj}[u]$
do RELAX(u, v, w)

Note: vertices in Q are in non-decreasing order of d , and there are only two values d , and $d+1$ in Q .

BFS(G, s)

1. For each vertex $u \in V$,
 $\text{color}[u] = W$; $d[u] = \infty$;
 $\pi[u] = \text{NIL}$; $\text{color}[s] = G$; $d[s] = 0$;
2. $Q = \{s\}$.
3. while Q is nonempty
 $u = \text{dequeue}(Q)$
for each $v \in \text{adj}[u]$,
if($\text{color}[v] == W$), do
 $\text{color}[v] = G$;
 $d[v] = d[u] + 1$;
 $\pi[v] = u$;
 $\text{enqueue}(Q, v)$;
 $\text{color}[u] = B$

Compared to PRIM

DIJKSTRA(G, w, s)

```
1. for each  $u \in V[G]$ 
2.   do  $d[u] \leftarrow \infty$ 
3.      $\pi[u] \leftarrow \text{NIL}$ 
4.  $d[s] \leftarrow 0$ 
5.  $s \leftarrow \emptyset$ 
6.  $Q \leftarrow V[G]$ 
7. while  $Q \neq \emptyset$ 
8. do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
9.  $s \leftarrow s \cup \{u\}$ 
10. for each  $v \in \text{Adj}[u]$ 
11. do if  $v \in Q$  and
       $d[u] + w(u, v) < d[v]$ 
12. then  $\pi[v] \leftarrow u$ 
13.      $d[v] \leftarrow d[u] + w(u, v)$ 
```

MST-PRIM(G, w, r)

```
1 for each  $u \in V[G]$ 
2   do  $key[u] \leftarrow \infty$ 
3      $\pi[u] \leftarrow \text{NIL}$ 
4  $key[r] \leftarrow 0$ 
5  $Q \leftarrow V[G]$ 
6 while  $Q \neq \emptyset$ 
7   do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
8     for each  $v \in \text{Adj}[u]$ 
9       do if  $v \in Q$  and  $w(u, v) < key[v]$ 
10         then  $\pi[v] \leftarrow u$ 
11            $key[v] \leftarrow w(u, v)$ 
```

Linear Programming

- Linear programming problem:

- Input: matrix A_{mn} , vector b_m , and vector c_n .

- Output: vector x_n , such that

maximize cx

subject to $Ax \leq b$.

- Notes

- This problem has polynomial time solution.

- Many problems can be converted into LP.

An Instance

单位产品所需原料数量（公斤）	产品Q1	产品Q2	产品Q3	原料供用量（公斤/日）
原料P1	2	3	0	1500
原料P2	0	2	4	800
原料P3	3	2	5	2000
单位产品的利润（千元）	3	5	4	

LP

$$\mathbf{max} \quad 3x_1 + 5x_2 + 4x_3$$

s.t.

$$2x_1 + 3x_2 \leq 1500$$

$$2x_2 + 4x_3 \leq 800$$

$$3x_1 + 2x_2 + 5x_3 \leq 2000$$

$$x_1, x_2, x_3 \geq 0$$

Feasibility problem of LP

- Feasible solution: x_n , subject to $Ax \leq b$.
 - Feasibility problem of LP:
 - Given Matrix A $m \times n$, b m -vector
 - Output: either a feasible solution x if one exists, or a judgment that no feasible solution exists.
-

Systems of difference constraints

- Each row of A contains exactly one “1” and one “-1”. All other elements are “0”.
- Each row is a difference constraint:
 - $x_j - x_i \leq b_k, 1 \leq k \leq m.$

- An example

$$\begin{pmatrix} 1 & -1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & 1 & 0 & 0 & -1 \\ -1 & 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \leq \begin{pmatrix} 0 \\ -1 \\ 1 \\ 5 \\ 4 \\ -1 \\ -3 \\ -3 \end{pmatrix}$$

$$\begin{array}{l} x_1 - x_2 \leq 0 \\ x_1 - x_5 \leq -1 \\ x_2 - x_5 \leq 1 \\ x_3 - x_1 \leq 5 \\ x_4 - x_1 \leq 4 \\ x_4 - x_3 \leq -1 \\ x_5 - x_3 \leq -3 \\ x_5 - x_4 \leq -3 \end{array}$$

Solution $x = (-5, -3, 0, -1, -4)$; $x' = (-5+d, -3+d, 0+d, -1+d, -4+d)$
 $(0, 2, 5, 4, 1)$

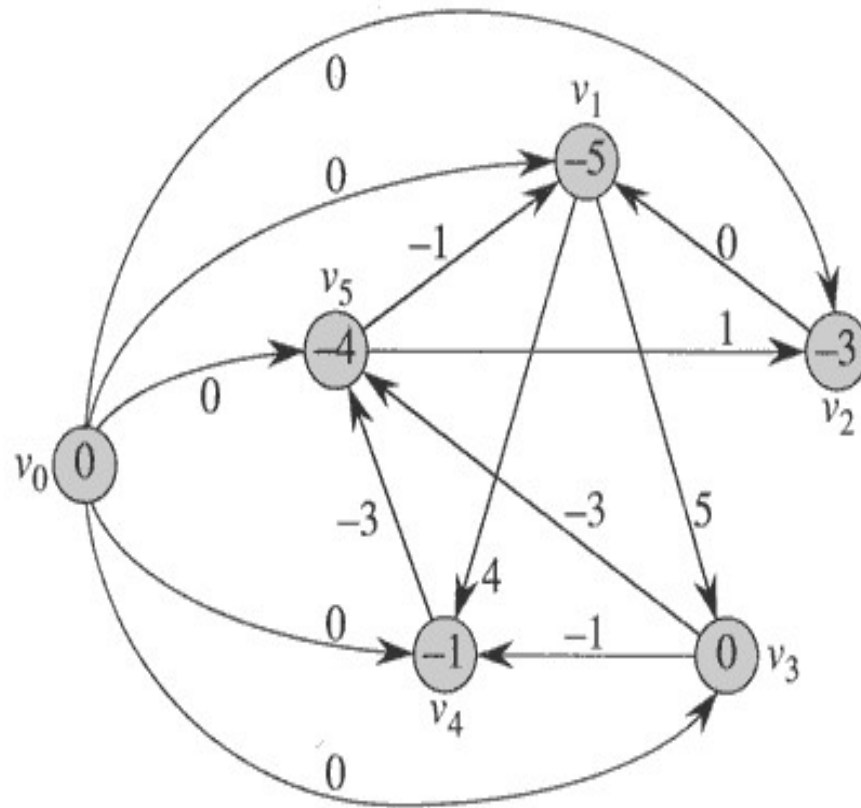
Relations between solutions to DC

- Lemma 24.8: Let $x=(x_1, x_2, \dots, x_n)$ be a solution to a system $Ax \leq b$ of difference constraints, and let d be any constant. Then $x+d=(x_1+d, \dots, x_n+d)$ is a solution to $Ax \leq b$ as well.
 - Proof. Easy(Book236)
-

Constraints graph

- Given a system $Ax \leq b$ of difference constraints
 - The constraint graph is a weighted, directed graph $G=(V, E; w)$, where:
 - $V = \{v_0, v_1, \dots, v_n\}$ // v_0 is an extra vertex
 - $E = \{(v_i, v_j) : x_j - x_i \leq b_k \text{ is a constraint}\} \cup \{(v_0, v_1), (v_0, v_2), \dots, (v_0, v_n)\}$. // each vertex is reachable from v_0
 - $w(v_i, v_j) = b_k$, if $x_j - x_i \leq b_k$ is a constraint;
 - $w(v_0, v_i) = 0$.
-

Constraint graph of example



$$\begin{pmatrix} 1 & -1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & 1 & 0 & 0 & -1 \\ -1 & 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \leq \begin{pmatrix} 0 \\ -1 \\ 1 \\ 5 \\ 4 \\ -1 \\ -3 \\ -3 \end{pmatrix}$$

Figure 24.8 The constraint graph corresponding to the system (24.3)–(24.10) of difference constraints. The value of $\delta(v_0, v_i)$ is shown in each vertex v_i . A feasible solution to the system is $x = (-5, -3, 0, -1, -4)$.

Solve Difference Constraints by SSSP in constraint graph

- Thoerem24.9: Given a system $Ax \leq b$ of difference constraints, let $G=(V, E)$ be the corresponding constraint graph.

- If G contains no negative-weight cycles, then

$$x = (\delta(v_0, v_1), \delta(v_0, v_2), \dots, \delta(v_0, v_n))$$

is a feasible solution for the system.

- If G contains a negative-weight cycle, then there is no feasible solution for the system.
-

Correctness of Theorem 24.9

- Proof. (\Rightarrow) If G does not contain negative-weight cycles. Then shortest path is well defined. And since there is an edge from v_0 to each vertex, **all shortest paths from v_0 is finite**. Consider any edge (v_i, v_j) (any constraint). By the triangle inequality, $\delta(v_0, v_j) \leq \delta(v_0, v_i) + w(v_i, v_j)$ or, equivalently, $\delta(v_0, v_j) - \delta(v_0, v_i) \leq w(v_i, v_j)$. Thus, letting $x_i = \delta(v_0, v_i)$ and $x_j = \delta(v_0, v_j)$ satisfies the difference constraint $x_j - x_i \leq b_k = w(v_i, v_j)$ that corresponds to edge (v_i, v_j) .
-

Proof continued

- If G contains a negative-weight cycle $C = \langle v_1, v_2, \dots, v_k \rangle$, where $v_1 = v_k$. We will show that there is no feasible solution. Cycle C corresponds to the following difference constraints: $x_2 - x_1 \leq w(v_1, v_2)$; $x_3 - x_2 \leq w(v_2, v_3)$; \dots , $x_k - x_{k-1} \leq w(v_{k-1}, v_k)$, $x_1 - x_k \leq w(v_k, v_1)$. If there is a feasible solution x , it will satisfy all these k inequalities. It also satisfies the summation of the inequalities, i.e., $0 \leq w(C)$, a contradiction to C is negative-weighted.
-

Property of DC

- Observation(According to Theorem24.9):
 - A difference constraints system has a solution
 \Leftrightarrow constraint graph contains no negative cycle, and
 $\mathbf{x}=(\delta(v_0,v_1), \delta(v_0,v_2), \dots, \delta(v_0,v_n))$ is a solution.
 - A difference constraints system has no solution \Leftrightarrow
constraint graph contains negative cycles.
-

Bellman-Ford solution to DC

Algorithm:

- Compute constraint graph G ;
 - Run Bellman-Ford on constraint graph G ;
 - If Bellman-Ford returns True, then $\mathbf{x}=(\delta(v_0, v_1), \dots, \delta(v_0, v_n))$ is a solution to DC; if Bellman-Ford returns false, DC has no solution.
-

Conclusion

- Relaxation and its properties
 - Bellman-Ford algorithm
 - Single-source shortest paths in DAG
 - Dijkstra's algorithm
 - Difference constraints
-

Homework

- 24.3-2
- 24.3-3



Next Class

- All-pairs shortest paths problem



实验五

- 给定一个有向加权完全图，每条边的权重为正，求最小的TSP圈。
 - TSP圈：经过所有点，且每个点仅经过一次的圈。
 - 方法自选，动态规划、分支定界、枚举都可以。