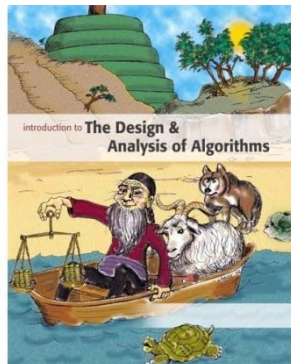




Introduction to

Algorithm Design and Analysis

[14] Minimum Spanning Tree



Yu Huang

<http://cs.nju.edu.cn/yuhuang>
Institute of Computer Software
Nanjing University



In the last class...

- **Undirected and Symmetric Digraph**
 - DFS skeleton
- **Biconnected Components**
 - Articulation point
 - Bridge
- **Other undirected graph problems**
 - Orientation for undirected graphs
 - MST based on graph traversal



当年我们学贪心算法

定义 7 一个图论算法的计算量 $f(v, \varepsilon) = O(P(v, \varepsilon))$ 时, 则称此算法为有效算法或好算法, 其中 $P(v, \varepsilon)$ 是某个多项式, v 与 ε 分别是图的顶点数与边数.

Dijkstra 算法 (u, v 不相邻时, $w(uv) = \infty$)

(1) 令 $l(u_0) = 0$; $l(v) = \infty, v \neq u_0$; $S_0 = \{u_0\}, i = 0$.

(2) 对每一个 $v \in \bar{S}_i$ (\bar{S}_i 指 S_i 以外的顶所成之集合), 用 $\min\{l(v), l(u_i) + w(u_i, v)\}$ 代替 $l(v)$; 设 u_{i+1} 是使 $l(v)$ 取最小值的 \bar{S}_i 中的顶, 令 $S_{i+1} = S_i \cup \{u_{i+1}\}$;

(3) 若 $i = v - 1$, 止; 若 $i < v - 1$, 用 $i + 1$ 代替 i , 转(2).

由上述算法知:

(1) S_i 中各顶标 $l(u)$ 即为 u_0 到 u 的距离. 又因 $v < \infty$, 故有限步之后, $V(G)$ 中每一顶都标志了与 u_0 的距离, 从而可以找到各顶到 u_0 的最短轨.

(2) Dijkstra 算法的时间复杂度 $f(v, \varepsilon) = O(v^2)$, 所以是有效算法.



Greedy Strategy

- **Optimization Problem**
- **Greedy Strategy**
- **MST Problem**
 - Prim's Algorithm
 - Kruskal's Algorithm
- **Single-Source Shortest Path Problem**
 - Dijkstra's Algorithm



Greedy Strategy for Optimization Problems

- **Coin change Problem**
 - [candidates] A finite set of coins, of 1, 5, 10 and 25 units, with enough number for each value
 - [constraints] Pay an exact amount by a selected set of coins
 - [optimization] a smallest possible number of coins in the selected set
- **Solution by greedy strategy**
 - For each selection, choose the highest-valued coin as possible.



Greedy Fails Sometimes

We have to pay 15 in total

- If the available types of coins are $\{1,5,12\}$
 - The greedy choice is $\{12,1,1,1\}$
 - But the smallest set of coins is $\{5,5,5\}$
- If the available types of coins are $\{1,5,10,25\}$
 - The greedy choice is always correct



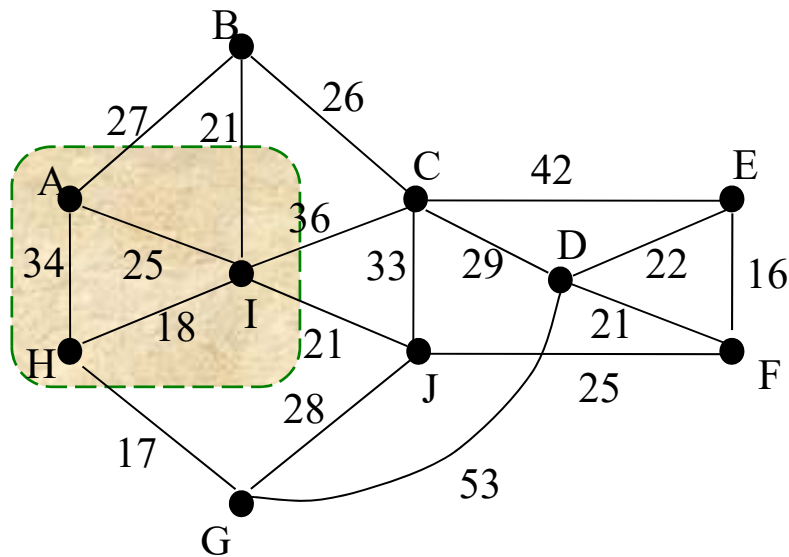
Greedy Strategy

- Expanding the partial solution **step by step**.
- In each step, a selection is made from a set of candidates. The choice made **must** be:
 - [Feasible] it has to satisfy the problem's constraints
 - [Locally optimal] it has to be the best local choice among all feasible choices on the step
 - [Irrevocable] the choice cannot be revoked in subsequent steps

```
set greedy(set candidate)
  set S=∅;
  while not solution(S) and candidate≠∅
    select locally optimizing x from candidate;
    candidate=candidate-{x};
    if feasible(x) then S=S∪{x};
  if solution(S) then return S
  else return ("no solution")
```



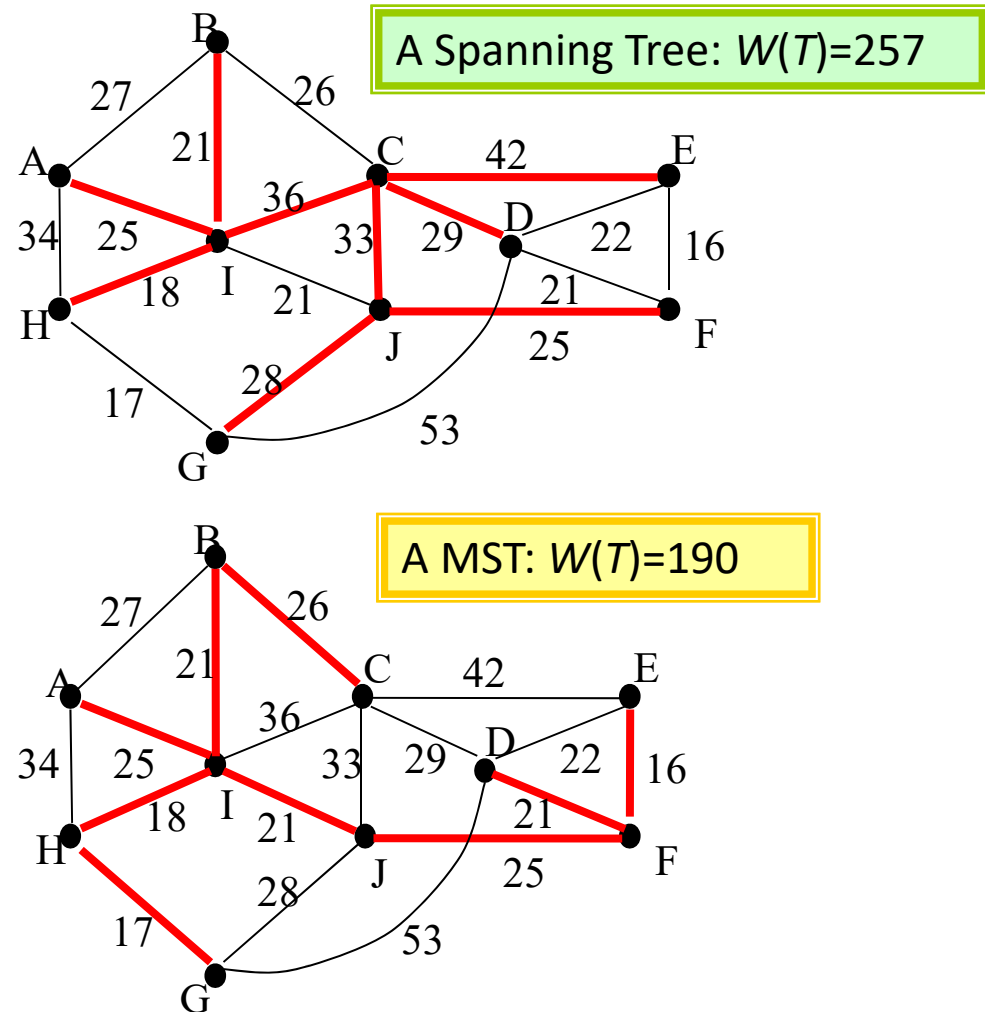
Weighted Graph and MST



A weighted graph

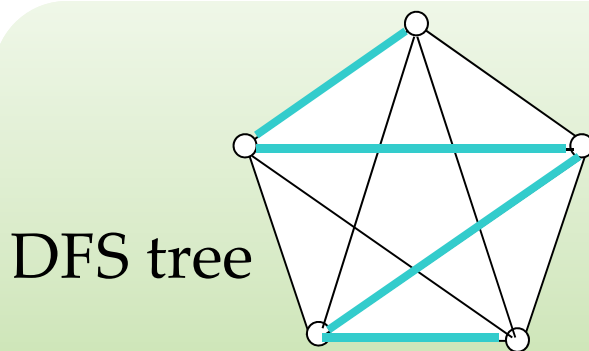
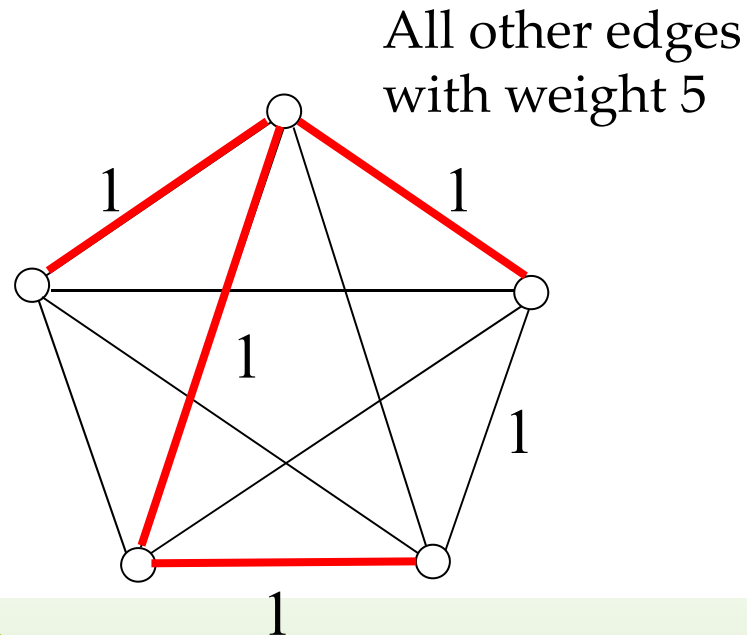
The nearest neighbor of vertex **I** is **H**

The nearest neighbor of shaded
subset of vertex is **G**

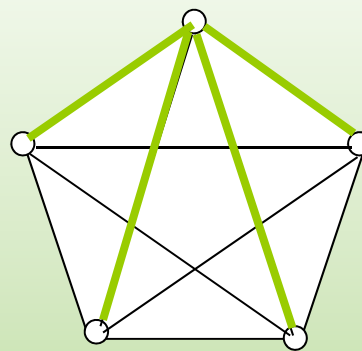


Graph Traversal and MST

There are cases that graph traversal tree **cannot** be minimum spanning tree, with the vertices explored in any order.



DFS tree



BFS tree

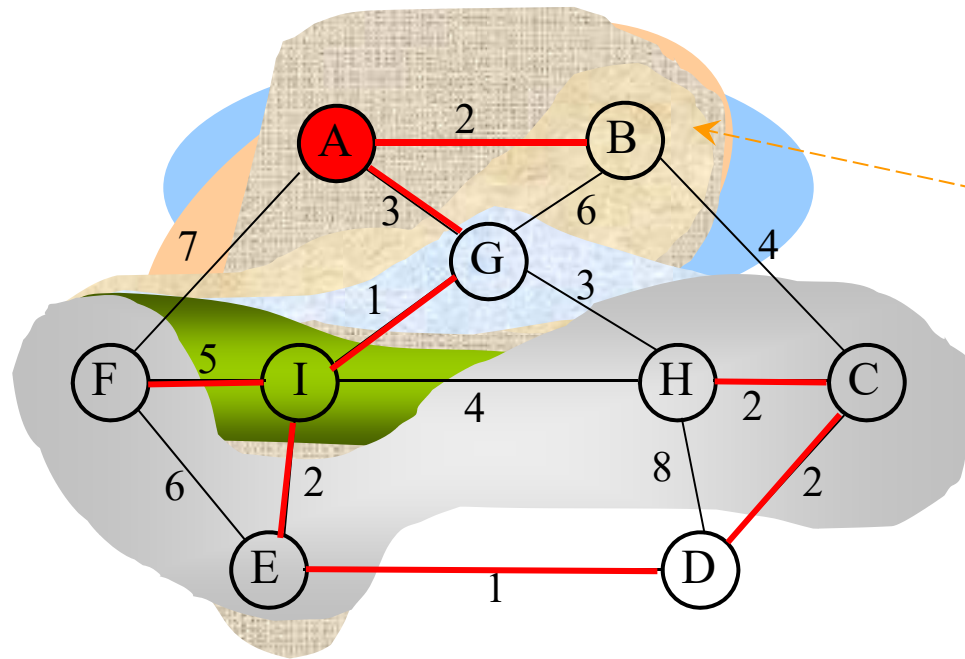
in any ordering of vertex

Greedy Algorithms for MST

- **Prim's algorithm:**
 - Difficult selecting: “best local optimization means **no cycle and small weight under limitation**.”
 - Easy checking: doing nothing
- **Kruskal's algorithm:**
 - Easy selecting: smallest in primitive meaning
 - Difficult checking: **no cycle**



Prim's Algorithm

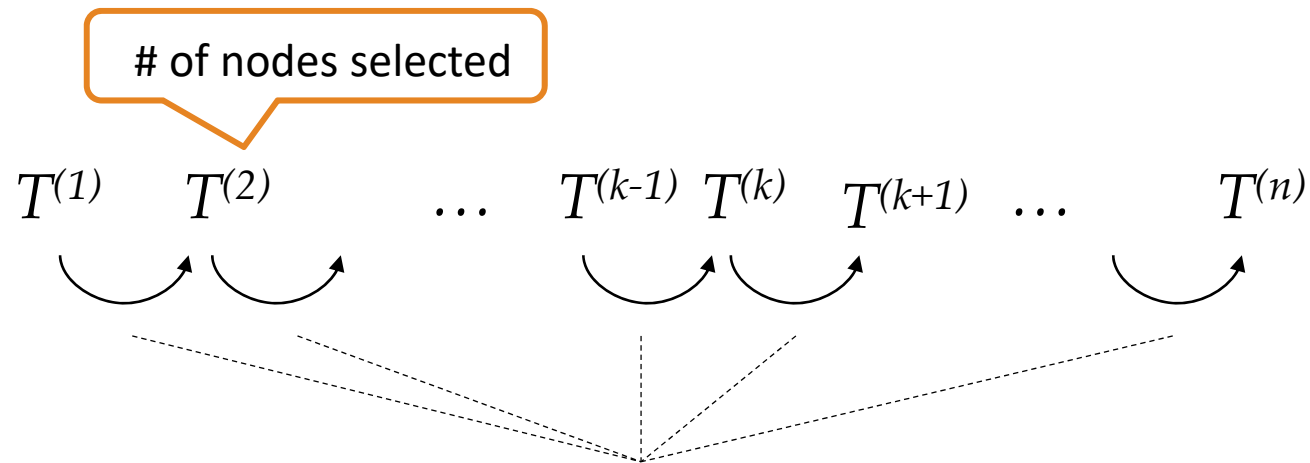


Greedy strategy:
For each set of fringe vertex,
select the edge with the
minimal weight, that is,
local optimal.

edges included in the MST



Correctness: How to Prove



Invariance: MST {

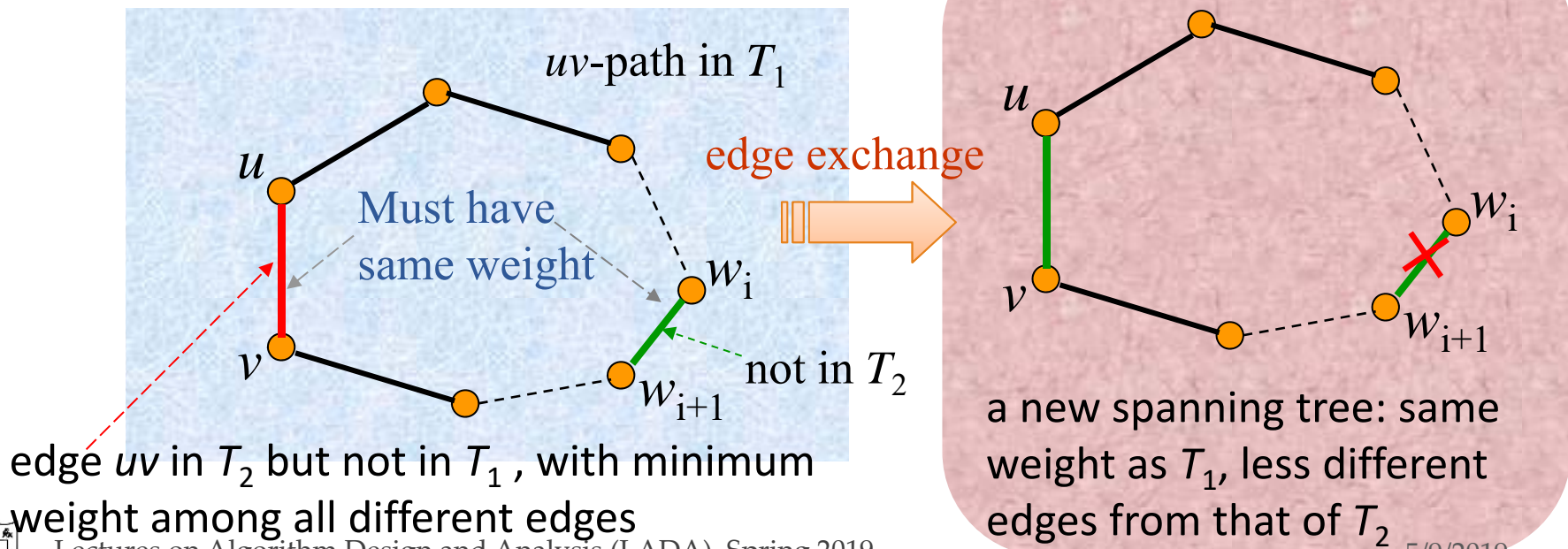
- Spanning tree
- Min weight

Definition transformation



Minimum Spanning Tree Property

- A spanning tree T of a connected, weighted graph has MST property if and only if for any non-tree edge uv , $T \cup \{uv\}$ contain a cycle in which uv is **one of** the maximum-weight edge.
- All the spanning trees having MST property have the same weight.



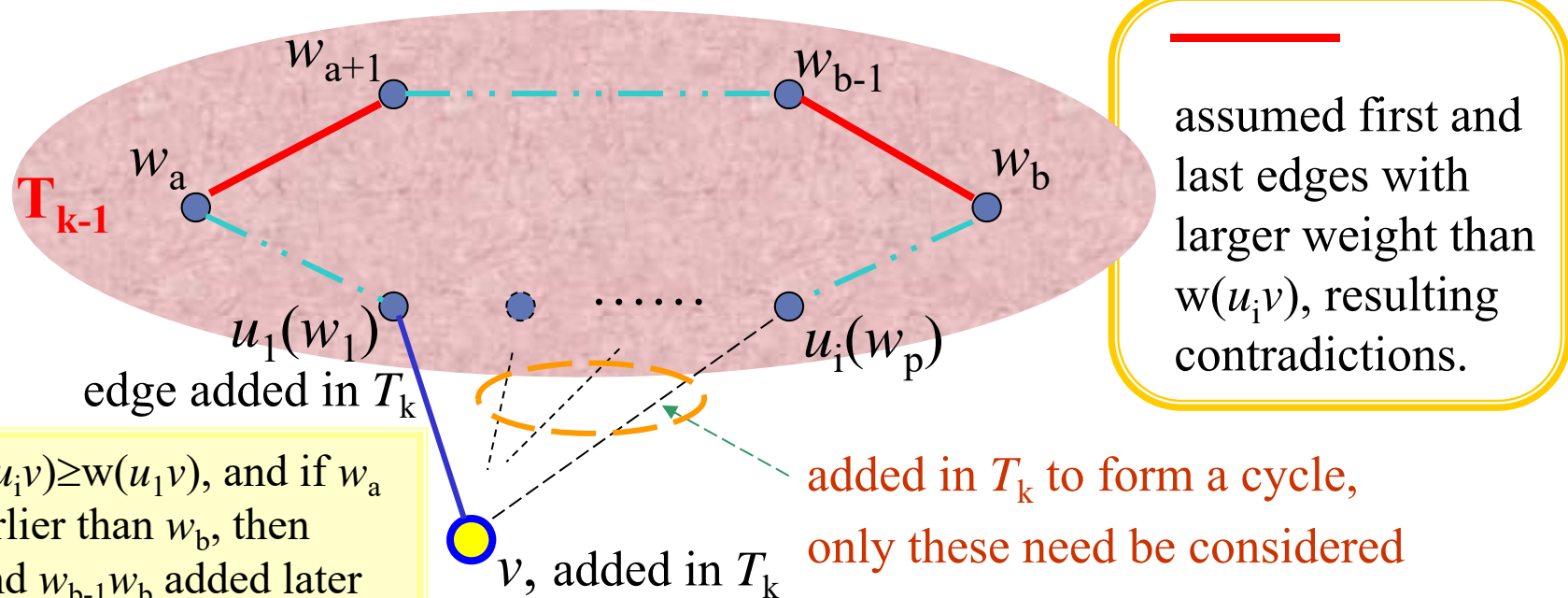
MST Property and Minimum Spanning Tree

- In a connected, weighted graph $G=(V,E,W)$, a tree T is a minimum spanning tree if and only if T has the MST property.
- **Proof**
 - \Rightarrow For a minimum spanning tree T , if it doesn't have MST property. So, there is a non-tree edge uv , and $T \cup \{uv\}$ contains an edge xy with weight larger than that of uv . Substituting uv for xy results in a spanning tree with less weight than T . Contradiction.
 - \Leftarrow As claimed above, any minimum spanning tree has the MST property. Since T has MST property, it has the same weight as any minimum spanning tree, i.e. T is a minimum spanning tree as well.



Correctness of Prim's Algorithm

- Let T_k be the tree constructed after the k^{th} step of Prim's algorithm is executed. Then T_k has the MST property in G_k , the subgraph of G induced by vertices of T_k .



Note: $w(u_1 v) \geq w(u_1 w_a)$, and if w_a added earlier than w_b , then $w_a w_{a+1}$ and $w_{b-1} w_b$ added later than any edges in $u_1 w_a$ -path, and v as well.

Key Issue in Implementation

- **Maintaining the set of fringe vertices**
 - Create the set and update it after each vertex is “selected” (*deleting* the vertex having been selected and *inserting* new fringe vertices)
 - Easy to decide the vertex with “highest priority”
 - Changing the priority of the vertices (*decreasing key*).
- **The choice: priority queue**



Implementation

Main Procedure

primMST(G, n)

Initialize the priority queue pq as empty;

Select vertex s to start the tree;

Set its candidate edge to $(-1, s, 0)$;

insert($pq, s, 0$);

while (pq is not empty)

$v = \text{getMin}(pq)$; $\text{deleteMin}(pq)$;

add the candidate edge of v to the tree;

updateFringe(pq, G, v);

return

$\text{getMin}(pq)$ always be the vertex with the smallest key in the fringe set.

ADT operation executions:

insert, **getMin**, **deleteMin**: n times

decreaseKey: m times

Updating the Queue

updateFringe(pq, G, v)

For all vertices w adjacent to v *// 2m loops*

$\text{newWgt} = w(v, w)$;

if $w.\text{status}$ is unseen **then**

Set its candidate edge to (v, w, newWgt) ;

insert(pq, w, newWgt)

else

if $\text{newWgt} < \text{getPriority}(pq, w)$

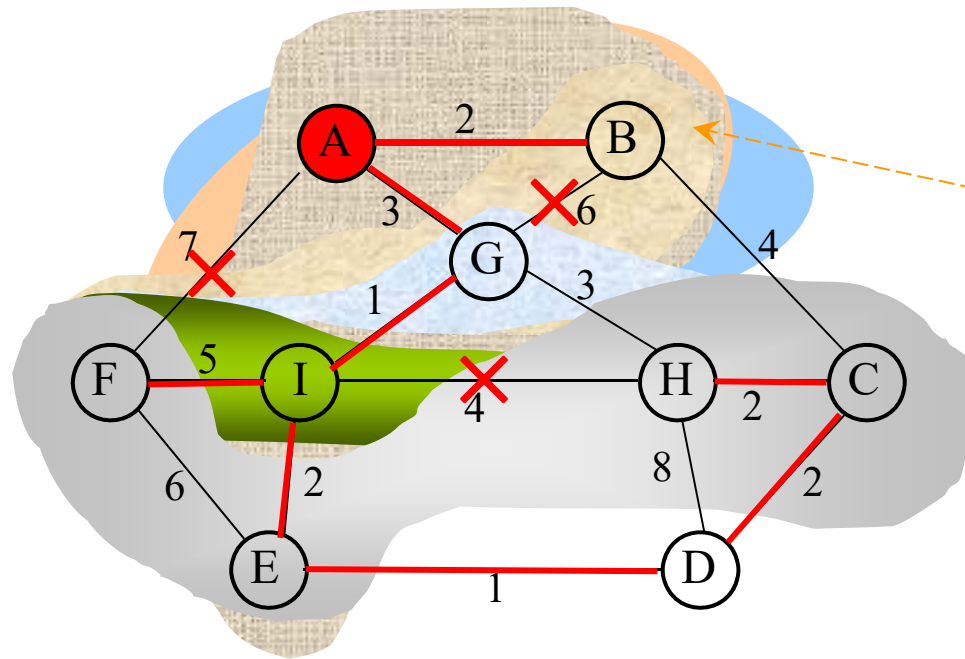
Revise its candidate edge to (v, w, newWgt) ;

decreaseKey(pq, w, newWgt)

return



Prim's Algorithm



Greedy strategy:
For each set of fringe vertex,
select the edge with the
minimal weight, that is,
local optimal.

edges included in the MST

Complexity

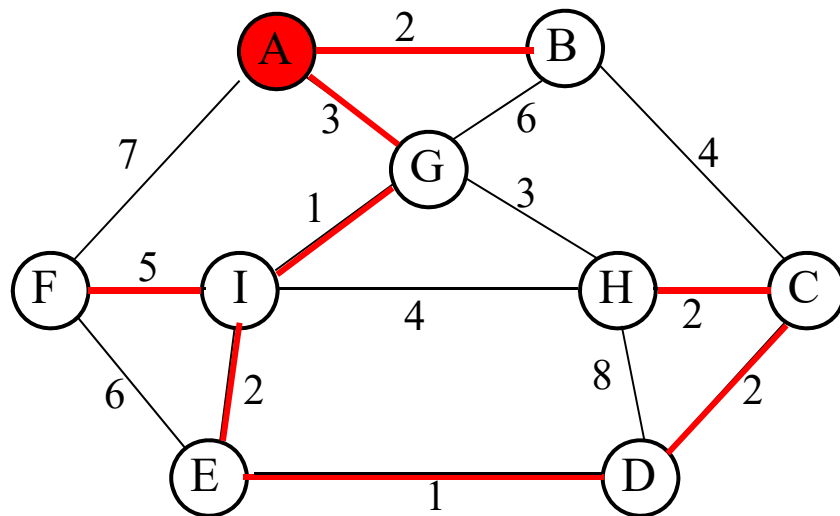
- Operations on ADT priority queue: (for a graph with n vertices and m edges)
 - insert: n ; getMin: n ; deleteMin: n ;
 - decreaseKey: m (appears in $2m$ loops, but execute at most m)
- So,

$$T(n,m) = O(nT(\text{getMin}) + nT(\text{deleteMin} + \text{insert}) + mT(\text{decreaseKey}))$$

- Implementing priority queue using array, we can get $\Theta(n^2 + m)$



Kruskal's Algorithm



edges included in the MST

Also Greedy strategy:
From the set of edges not yet included in the partially built MST, select the edge with the minimal weight, that is, local optimal, in another sense.

Key Issue in Implementation

- How to know an insertion of edge will result in a cycle *efficiently*?
- For correctness: the two endpoints of the selected edge *can not* be in the same connected components.
- For the efficiency: connected components are implemented as dynamic equivalence classes using union-find.



Kruskal's Algorithm: the Procedure

- `kruskalMST(G,n,F)` //outline
- `int count;`
- Build a minimizing priority queue, `pq`, of edges of `G`, prioritized by weight.
- Initialize a Union-Find structure, `sets`, in which each vertex of `G` is in its own set.
-
- `F = ϕ ;`
- `while (isEmpty(pq) == false)`
- `vwEdge = getMin(pq);`
- `deleteMin(pq);`
- `int vSet = find(sets, vwEdge.from);`
- `int wSet = find(sets, vwEdge.to);`
- `if (vSet \neq wSet)`
- `Add vwEdge to F;`
- `union(sets, vSet, wSet)`
- `return`

Simply sorting, the cost will be $\Theta(m \log m)$



Prim vs. Kruskal

- **Lower bound for MST**
 - For a correct MST, each edge in the graph should be examined at least once.
 - So, the lower bound is $\Omega(m)$
- **$\Theta(n^2+m)$ and $\Theta(m\log m)$, which is better?**
 - Generally speaking, depends on the density of edge of the graph.



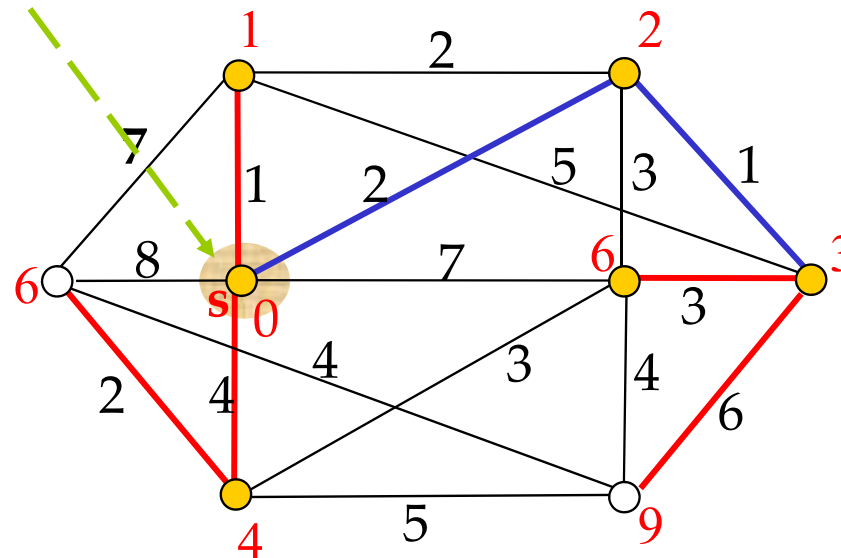
Single Source Shortest Paths

The single source

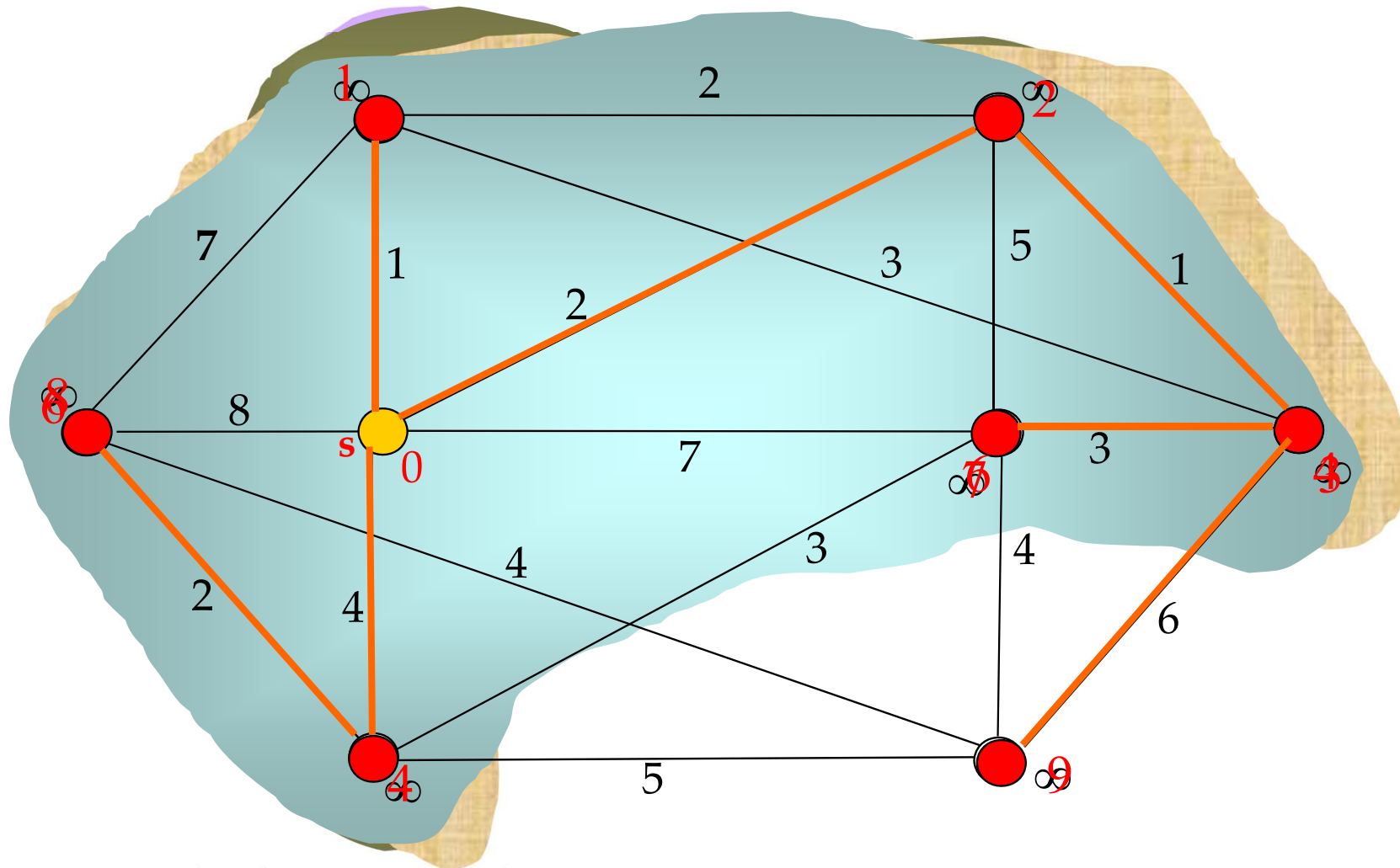
Red labels on each vertex is the length of the shortest path from s to the vertex.

Note:

The shortest $[0, 3]$ -path doesn't contain the shortest edge leaving s , the edge $[0, 1]$



Dijkstra's Algorithm



Thank you!

Q & A

Yu Huang

<http://cs.nju.edu.cn/yuhuang>

