

Hanoi University of Science and Technology  
School of Information and Communications Technology


# Data structures and Algorithms

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## Course outline

- Chapter 1. Fundamentals
- Chapter 2. Algorithmic paradigms
- Chapter 3. Basic data structures
- Chapter 4. Tree
- Chapter 5. Sorting
- Chapter 6. Searching
- Chapter 7. Graph**

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# Chapter 7. Graph

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## Shortest path representation

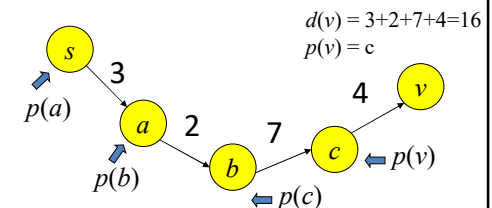
Shortest path algorithms works on 2 arrays:

- ✦  $d(v)$  = the length of shortest path from  $s$  to  $v$  that algorithm found so far (upper bound for the length of the shortest path from  $s$  to  $v$ ).
- ✦  $p(v)$  = a predecessor of  $v$  in this shortest path (used to back trace the path from  $s$  to  $v$ ).  $\delta(s,v) \leq d(v)$

### Initialization

```

for  $v \in V(G)$ 
do  $d[v] \leftarrow \infty$ 
    $p[v] \leftarrow \text{NULL}$ 
 $d[s] \leftarrow 0$ 
    
```



## Relaxation

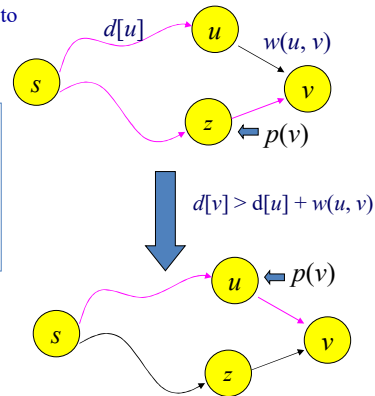
Building the shortest path from  $s$  to  $v$ :

Assume the current known shortest path from  $s$  to  $v$ :  $s \dots z \rightarrow v$

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

```

Relax( $u, v$ )
if ( $d[v] > d[u] + w(u, v)$ )
{
     $d[v] \leftarrow d[u] + w(u, v)$ 
     $p[v] \leftarrow u$ 
}
    
```



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## Properties of Relaxation

```

Relax( $u, v$ )
if ( $d[v] > d[u] + w(u, v)$ )
{
     $d[v] \leftarrow d[u] + w(u, v)$ 
     $p[v] \leftarrow u$ 
}
    
```

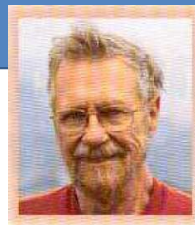
Shortest path algorithms differ in

- how many times they relax each edge, and
- the order in which they relax edges

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## Dijkstra Algorithm

- In case the weights on the edges are non-negative, the algorithm proposed by Dijkstra is more efficient than the Ford-Bellman algorithm.
- Algorithms are built by labeling vertices. The label of the vertices is initially temporary. At each iteration there is a temporary label that becomes a permanent label. If the label of a vertex  $u$  becomes fixed,  $d[u]$  gives us the length of the shortest path from the source  $s$  to  $u$ . Algorithm ends when the labels of all vertices become fixed.



Edsger W. Dijkstra  
(1930-2002)

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## Dijkstra algorithm

- Input:** A directed graph  $G=(V,E)$  and weight matrix  $w[u,v] \geq 0$  where  $u,v \in V$ , source vertex  $s \in V$ ;
  - $G$  does not have negative-weight cycle
- Output:** Each  $v \in V$ 
  - $d[v] = \delta(s, v)$ ; Length of the shortest path from  $s$  to  $v$
  - $p[v]$  - the predecessor of  $v$  in this shortest path from  $s$  to  $v$ .

Use greedy algorithm:

Maintain a set  $S$  of vertices for which we know the shortest path

At each iteration:

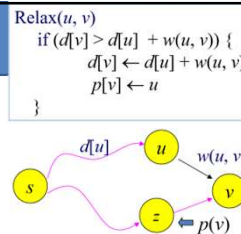
- grow  $S$  by one vertex, choosing shortest path through  $S$  to any other vertex not in  $S$
- If the cost from  $S$  to any other vertex has decreased, update it

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## Dijkstra algorithm

```

Dijkstra ( )
{
    for v in V \ s { // Initialize
        d[v] = w[s,v];
        p[v] = s;
    }
    d[s] = 0; p[s] = s;
    S = {s}; // S: the set of vertices with fixed label (shortest path from s to it has been found)
    T = V \ {s}; // T: the set of vertices with temporary label
    while (T != ∅) // Loop
    {
        Find vertex u in T satisfying d[u] = min{ d[z] : z in T };
        T = T \ {u}; S = S ∪ {u}; // Fixed label of vertex u
        for v in adj[u] and v in T // Assign new label to each vertex v of T if necessary (if value d[v] is decreased)
            Relax(u, v)
    }
}
    
```



Use greedy algorithm:

Maintain a set  $S$  of vertices for which we know the shortest path

At each iteration:

- grow  $S$  by one vertex, choosing shortest path through  $S$  to any other vertex not in  $S$
- If the cost from  $S$  to any other vertex has decreased, update it

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## Dijkstra algorithm

```

void Dijkstra ( )
{
    for v in V \ s // Initialize
    {
        d[v] = w[s,v];
        p[v] = s;
    }
    d[s] = 0; p[s] = s; S = {s};
    T = V \ {s};
    while (T != ∅) // Loop
    {
        Find vertex u in T satisfying d[u] = min{ d[z] : z in T };
        T = T \ {u}; S = S ∪ {u};
        for v in adj[u] and v in T
        {
            if (d[v] > d[u] + w[u,v])
            {
                d[v] = d[u] + w[u,v];
                p[v] = u;
            }
        }
    }
}
    
```

$O(|V|^2)$

$O(|E|)$  for whole loop, as the adjacent list each vertex of the graph will be traversed exactly once

**The computation time:  $O(|V|^2 + |E|)$**

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## Dijkstra algorithm

- Comment:** If only need to find the shortest path from  $s$  to  $t$  then the algorithm could stop when  $t$  has fixed label ( $t \in S$ ).

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## (1) Direct implement

```

Dijkstra_Table(G, s)
1. for v in V \ s do {
2.     d[v] ← infinity;
3.     found[v] ← FALSE;
4. }
5. d[s] ← 0; found[s] = TRUE;
   // s is source vertex
6. for i = 1 to |V|-1 do {
7.     u ← the vertex with min d among all
       vertices k having found[k] = false;
8.     found[u]=TRUE;
9.     for (v in Adj(u) && !found[v]) do
10.        if (d[v] > d[u] + c[u, v]) {
11.            d[v] = d[u] + c[u, v];
12.            p[v] = u;
13.        }
14. }
    
```

Computation time **Dijkstra\_Table(G, s):**  $O(|V|^2 + |E|)$ .

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```

INF = int(1e9)
def input_data():
    global V, E, s, t, Adj
    V, E = map(int, input().split())
    Adj = [[] for _ in range(V+1)]
    for _ in range(E):
        u, v, weight = map(int, input().split())
        Adj[u].append((v, weight))
    s, t = map(int, input().split())
def dijkstra_table():
    INF = int(1e9)
    V, E, s, t, Adj = input_data()
    d = [INF] * (V+1)
    found = [False] * (V+1)
    d[s] = 0
    found[s] = True
    for i in range(1, V):
        u = None
        for v in range(1, V+1):
            if not found[v] and d[v] < d[u]:
                u = v
        found[u] = True
        for (v, weight) in Adj[u]:
            if d[v] > d[u] + weight:
                d[v] = d[u] + weight
    return d[t]
    
```

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## (2) Implement Dijkstra by using priority queue

For each iteration of the algorithm we need to find a vertex with minimum  $d$ , to implement it more efficiently, we use min priority queue

```

Dijkstra_Table(G, s)
1. for u in V do {
2.     d[u] ← infinity;
3.     found[u] ← FALSE;
4. }
5. d[s] ← 0;
   // s is source vertex
6. for i = 1 to |V|-1 do {
7.     u ← the vertex with min d among all
       vertices k having found[k] = false;
8.     found[u]=TRUE;
9.     for (v in Adj(u) && !found[v]) do
10.        if (d[v] > d[u] + c[u, v]) {
11.            d[v] = d[u] + c[u, v];
12.            p[v] = u;
13.        }
14. }
    
```

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## (2) Implement Dijkstra by using priority queue

```

Dijkstra_Heap(G, s)
1. PQ = []
2. for u ∈ V do {
3.   d[u] = infinity;
4.   found[u] = FALSE;
5. }
6. d[s] ← 0; // s is source vertex
7. PQ.enqueue(s, d[s]);
8. for i = 1 to |V|-1 do {
9.   u ← PQ.dequeue(); // get vertex with min d
10.  found[u] = TRUE;
11.  for (v ∈ Adj(u) && !found[v]) do
12.    if (d[v] > d[u] + c[u, v]) {
13.      d[v] = d[u] + c[u, v];
14.      p[v] = u;
15.      PQ.Decrease_Key(v, d[v]);
16.    }
17. }

```

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## (2) Implement Dijkstra by using priority queue

**The computation time:**  
 $O((|E|+|V|)\log|V|)$

```

Dijkstra_Heap(G, s)
1. PQ = []
2. for u ∈ V do {
3.   d[u] = infinity;
4.   found[u] = FALSE;
5. }
6. d[s] ← 0;
7. PQ.enqueue(s, d[s]); // s is source vertex → O(log|V|)
8. for i = 1 to |V|-1 do {
9.   u ← PQ.dequeue(); // get vertex with min d → O(log|V|)
10.  found[u] = TRUE;
11.  for (v ∈ Adj(u) && !found[v]) do
12.    if (d[v] > d[u] + c[u, v]) {
13.      d[v] = d[u] + c[u, v];
14.      p[v] = u;
15.      PQ.Decrease_Key(v, d[v]); → O(log|V|)
16.    }
17. }

```

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2. **Kruskal algorithm**

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## Kruskal algorithm

```

Generic-MST(G, c)
T = ∅
// T is the subset edges of a minimum spanning tree
while T is not the spanning tree do
  Finding edge (u, v) is "safe" edge for T
  T = T ∪ {(u, v)}
return T

```



Joseph Kruskal  
(1928 - ~)

### Kruskal algorithm:

- ❖ *T* is forest (empty).
- ❖ The "safe" edge included in *T* at each iteration is the edge with smallest weight among edges connecting its connected components.

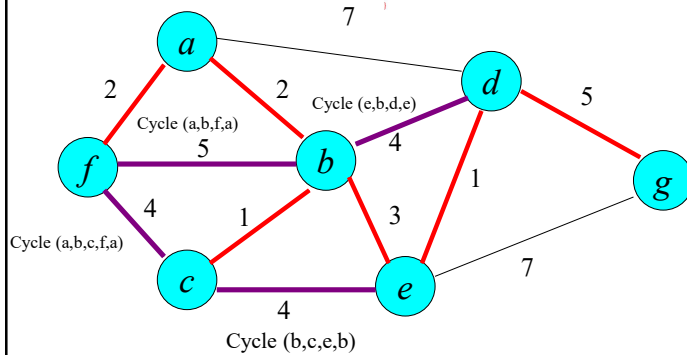
```

void Kruskal ( )
{
  Sort m edges of the graph e1, e2, . . . , em in ascending order of weight;
  T = ∅; // T: set of edges of the minimum spanning tree
  for (i = 1; i <= m; i++)
    if (T ∪ {ei} does not contain cycle)
      T = T ∪ {ei};
}

```

## Kruskal algorithm: an example

```
void Kruskal ( )
{
    Sort m edges of the graph  $e_1, e_2, \dots, e_m$  in ascending order of weight;
     $T = \emptyset$ ; //  $T$ : set of edges of the minimum spanning tree
    for (i = 1; i <= m; i++)
        if (  $T \cup \{e_i\}$  does not contain cycle )
             $T = T \cup \{e_i\}$ ;
}
```



Weight of the minimum spanning tree:  
 $2+2+1+3+1+5=14$

## Computation time

```
void Kruskal ( )
{
    Sort m edges of the graph  $e_1, e_2, \dots, e_m$  in ascending order of weight;
     $T = \emptyset$ ; //  $T$ : set of edges of the minimum spanning tree
    for (i = 1; i <= m; i++)
        if (  $T \cup \{e_i\}$  does not contain cycle )
             $T = T \cup \{e_i\}$ ;
}
```

- Step 1. Sort the edges in ascending order of weight
  - Could use heap sort/merge sort :  $O(m \log m)$
- Iterations: Each iteration we need to check whether  $T \cup \{e_i\}$  contains the cycle?
  - Could use DFS to check with time  $O(m+n)$ .
  - Total time:  $O(m(m+n))$

Computation time:  $O(m \log m + m(m+n))$  where  $n, m$  is the number of vertices and edges of the graph, respectively.

## Improved implementation:

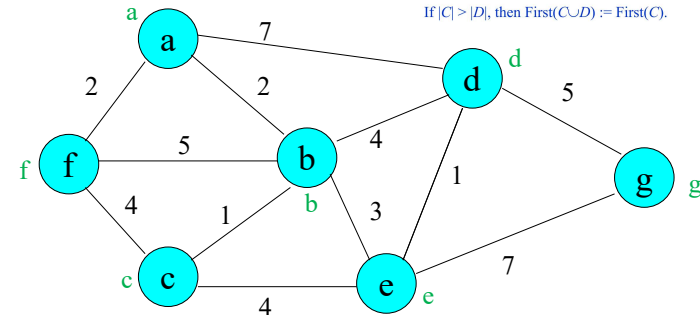
- Each connected component  $C$  of forest  $F$  is setup as a set.
- Denote  $\text{First}(C)$  be the first vertex in connected component  $C$ .
- Each vertex  $j$  in  $C$ , set  $\text{First}(j) = \text{First}(C) =$  first vertex in  $C$ .
- Note: Adding edge  $(i,j)$  to forest  $F$  creates cycle iff  $i$  and  $j$  belongs to the same connected component, it means  $\text{First}(i) = \text{First}(j)$ .
- When connecting connected components  $C$  and  $D$  together, we connect the **smaller** one (less number of vertices) to the **larger** one (more number of vertices):

If  $|C| > |D|$ , then  $\text{First}(C \cup D) := \text{First}(C)$ .

- ❖  $T$  is forest (empty).
- ❖ The "safe" edge included in  $T$  at each iteration is the edge with smallest weight among edges connecting its connected components.

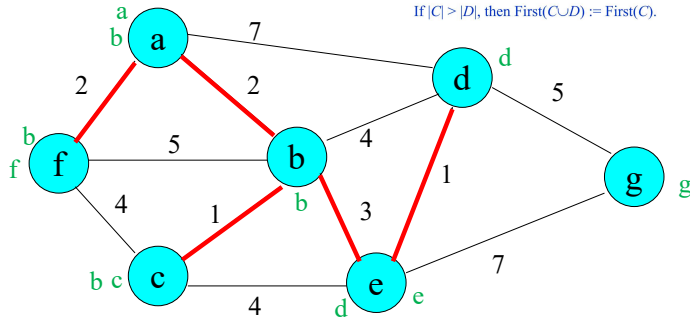
## Kruskal – Example

- Denote  $\text{First}(C)$  be the first vertex in connected component  $C$ .
- Each vertex  $j$  in  $C$ , set  $\text{First}(j) = \text{First}(C) =$  first vertex in  $C$ .
- Note: Adding edge  $(i,j)$  to forest  $F$  creates cycle iff  $i$  and  $j$  belongs to the same connected component, it means  $\text{First}(i) = \text{First}(j)$ .
- When connecting connected components  $C$  and  $D$  together, we connect the **smaller** one (less number of vertices) to the **larger** one (more number of vertices):  
 If  $|C| > |D|$ , then  $\text{First}(C \cup D) := \text{First}(C)$ .



## Kruskal – Example

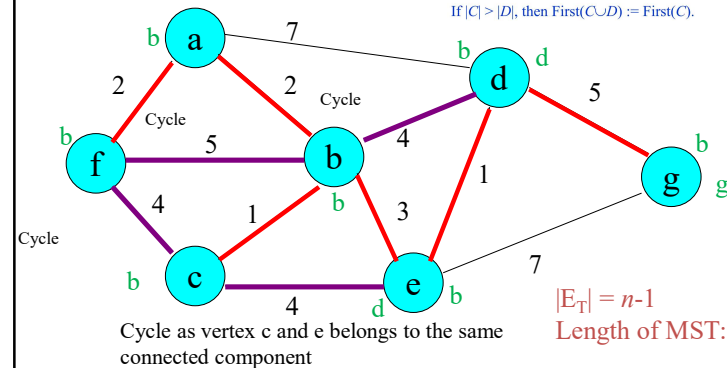
- Denote  $\text{First}(C)$  be the first vertex in connected component  $C$ .
- Each vertex  $j$  in  $C$ , set  $\text{First}(j) = \text{First}(C)$  = first vertex in  $C$ .
- Note: Adding edge  $(i, j)$  to forest  $F$  creates cycle **iff**  $i$  and  $j$  belongs to the same connected component, it means  $\text{First}(i) = \text{First}(j)$ .
- When connecting connected components  $C$  and  $D$  together, we connect the **smaller** one (less number of vertices) **to the larger** one (more number of vertices):  
If  $|C| > |D|$ , then  $\text{First}(C \cup D) := \text{First}(C)$ .



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## Kruskal – Example

- Denote  $\text{First}(C)$  be the first vertex in connected component  $C$ .
- Each vertex  $j$  in  $C$ , set  $\text{First}(j) = \text{First}(C)$  = first vertex in  $C$ .
- Note: Adding edge  $(i, j)$  to forest  $F$  creates cycle **iff**  $i$  and  $j$  belongs to the same connected component, it means  $\text{First}(i) = \text{First}(j)$ .
- When connecting connected components  $C$  and  $D$  together, we connect the **smaller** one (less number of vertices) **to the larger** one (more number of vertices):  
If  $|C| > |D|$ , then  $\text{First}(C \cup D) := \text{First}(C)$ .



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## Improved implementation: Computation time

- Time to determine whether 2 vertices  $i, j$  belong to the same connected component:  $\text{First}(i) = \text{First}(j) : O(1)$  for each  $i, j$ . There are  $m$  edges  $\rightarrow$  Total:  $O(m)$ .
- Time to connect 2 connected components  $S$  and  $Q$ , where  $|S| \geq |Q|$ :
  - $O(1)$  for each vertex of  $Q$  (the one with smaller number of vertices)
  - Each vertex  $i$  in smaller connected component: connect **log n times as maximum**. (As the number vertices of connected component containing  $i$  is doubled after each connection.)

Total time to connect over all algorithm:  $O(n \log n)$ .

- Computation time:  
 $O(m \log m + m + n \log n)$ .

## Improved implementation: Computation time

```
void Kruskal ( )
{
    Sort m edges of the graph  $e_1, e_2, \dots, e_m$  in ascending order of weight;
     $T = \emptyset$ ; //  $T$ : set of edges of the minimum spanning tree
    for (i = 1; i <= m; i++)
        if (  $T \cup \{e_i\}$  does not contain cycle)
             $T = T \cup \{e_i\}$ ;
}
```

Time to connect 2 connected components  $S$  and  $Q$ , where  $|S| \geq |Q|$ :

- $O(1)$  for each vertex of  $Q$  (the one with smaller number of vertices)
- Each vertex  $i$  in smaller connected component: connect **log n times as maximum**. (As the number vertices of connected component containing  $i$  is doubled after each connection.)

$\rightarrow$  Total time to connect over all algorithm:  $O(n \log n)$ .

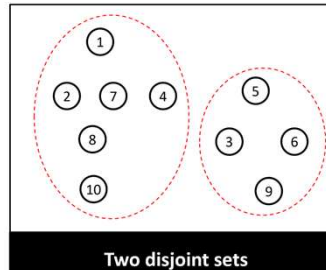
- Step 1. Sort the edges in ascending order of weight
  - Could use heap sort/merge sort :  $O(m \log m)$
- Iterations: Each iteration we need to check whether  $T \cup \{e_i = (x, y)\}$  contains the cycle?
  - Could use DFS to check with time  $O(m+n)$ .  $\rightarrow O(1)$ : check  $\text{First}(x) = \text{First}(y)$
  - Total time:  $O(m(m+n)) \rightarrow O(m)$

Computation time:  $O(m \log m + m(m+n))$  where  $n, m$  is the number of vertices and edges of the graph, respectively.  $\rightarrow O(m \log m + m + n \log n)$ .

## Disjoint set

- Disjoint set data structure is used to represent sets that don't have any elements in common. It supports 3 main operations:

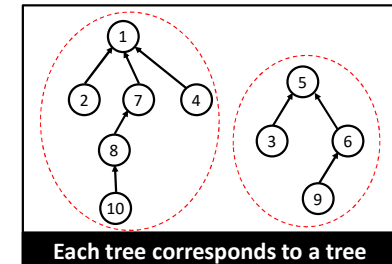
- makeSet(x)**: create a set containing only an element  $x$
- find(x)**: find the id of a set containing element  $x$
- union(a, b)**: merge the set having id= $a$  and the set having id= $b$  to a single set



## Disjoint set

- Each set is represented by a tree:
  - Each node in the tree corresponds to an element of the set.
  - The root of the tree is the id of the set.
  - Each node has exactly one parent (parent of the root is itself).

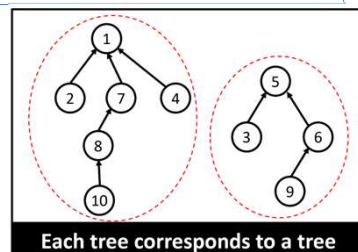
x	1	2	3	4	5	6	7	8	9	10
parent[x]	1	1	5	1	5	5	1	7	6	8



## Disjoint set

```

1. makeSet(x){ //Create a set containing only element x
2.     parent[x] = x;
3. }
4. find(x){ //return id of the set containing x
5.     while (x != parent[x]) do x = parent[x];
6.     return x;
7. }
```



## Disjoint set

```

1. makeSet(x){ //Create a set containing only element x
2.     parent[x] = x;
3. }
4. find(x){ //return id of the set containing x
5.     while (x != parent[x]) do x = parent[x];
6.     return x;
7. }
8. union(a, b) { //merge 2 sets: the set with id = a and the set with id = b
11.     if (a != b) then
12.         parent[a] = b; //make id of a's set be the id of b's set
```



## Kruskal: NOT use disjoint set

```

1. Kruskal(G=(V,E),w){
2.   E' = sort edges in ascending order of weight;
3.   Et = ∅
4.   for (u, v) in E' do {
5.     if Et ∪ (u, v) does not contain cycle then
6.       Et = Et ∪ (u, v);
7.     if (|Et| == |V| - 1) then break;
8.   }
9. }
10. if (|Et| < |V| - 1) then {
11.   print("ERROR: Graph G is not connected");
12.   return NULL;
13. }
14. else return (V, Et);
15. }
```

Not use Disjoint set

$$O(|E| \log |E| + |E|(|E| + |V|))$$

```

1. Kruskal(G=(V,E),w){
2.   for (x = 1; x <= |V|; x++) do makeSet(x);
3.   E' = sort edges in ascending order of weight;
4.   Et = ∅
5.   for (u, v) in E' do {
6.     idu = find(u);
7.     idv = find(v);
8.     if (idu != idv) then
9.       union(idu, idv);
10.    Et = Et ∪ (u, v);
11.    if (|Et| == |V| - 1) then break;
12.   }
13. }
14. if (|Et| < |V| - 1) then {
15.   print("ERROR: Graph G is not connected");
16.   return NULL;
17. }
18. else return (V, Et);
19. }
```

Use Disjoint set

What is the computation time if disjoint set is used ?

$$O(|V| + |E| \log |E| + |E||V|)$$

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## Kruskal: use disjoint set

```

1. makeSet(x) { //Create a set containing only element x
2.   parent[x] = x;
3. }
4. find(x) { //not compression
5.   while (x != parent[x]) do x = parent[x];
6.   return x;
7. }
8. union(a, b) { //not union by rank
9.   if (a != b) then
10.    parent[a] = b;
11. }
```

not optimize

$$O(|V| + |E| \log |E| + |E||V|)$$

```

1. Kruskal(G=(V,E),w){
2.   for (x = 1; x <= |V|; x++) do makeSet(x);
3.   E' = sort edges in ascending order of weight;
4.   Et = ∅
5.   for (u, v) in E' do {
6.     idu = find(u);
7.     idv = find(v);
8.     if (idu != idv) then
9.       union(idu, idv);
10.    Et = Et ∪ (u, v);
11.    if (|Et| == |V| - 1) then break;
12.   }
13. }
14. if (|Et| < |V| - 1) then {
15.   print("ERROR: Graph G is not connected");
16.   return NULL;
17. }
18. else return (V, Et);
19. }
```

If use union(a, b) not by rank and find(x) not compression:

- **find(x)**: could take  $O(|V|)$  in a worst-case scenario (a long, skinny tree).
- **union(a, b)**: Each union(a, b) operation takes  $O(1)$  because we just update the parent.

→ Kruskal: We call find(x) multiple times → the loop lines 5-13: the overall cost can be  $O(|E||V|)$ .

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## Disjoint set

- Path compression: to improve computation time of function find(x)

- When find(x) is called, root of the tree containing x is returned.

- Current implementation: The find(x) operation traverses up from x to find root.

```

4. find(x) { //return id of the set containing x
5.   while (x != parent[x]) do x = parent[x];
6.   return x;
7. }
```

- The idea of path compression is to make the found root as parent of x so that we don't have to traverse all intermediate nodes again. If x is root of a subtree, then path (to root) from all nodes under x also compresses. → It speeds up the data structure by **compressing the height** of the trees.

## Disjoint set

- Path compression: to improve computation time of function find(x)

```

1. find(x) { //return id of the set containing x
2.   while (x != parent[x]) do x = parent[x];
3.   return x;
4. }
```

↓ Path compression

```

find(x) {
  if (x != parent[x])
    parent[x] = find(parent[x]) #path compression
  return parent[x];
}
```

### Disjoint set: Path compression

```

find(x) {
  if (x != parent[x])
    parent[x] = find(parent[x])  #path compression
  return parent[x];
}
    
```

Set {0, 1,...,9} is represented as

Call find(3): traverse up and find 9 is id of the tree

When find(3) is terminated, path compression is already applied → we get directly link element 0 and 3 to 9

- The idea of path compression is to make the found root as parent of x so that we don't have to traverse all intermediate nodes again. If x is root of a subtree, then path (to root) from all nodes under x also compresses. → It speeds up the data structure by compressing the height of the trees.

### Disjoint set

- Reduce the size of each tree: to improve computation time of function union(x, y)
  - When merging two trees, we will join the tree with the lower height to the tree with the higher height.
  - We use the additional variable  $r[x]$  to record the height of a node x in the tree

Union 2 disjoint sets

### Disjoint set

```

1. union(a, b) { //merging 2 sets: the set with id=a and the set with id=b
2.   if (a != b) then
3.     parent[a] = b; //make id of a's set be the id of b's set
    
```

```

1. union(a, b) { //merging 2 sets: the set with id=a and the set with id=b
2.   if (a != b) then
3.     if (r[a] > r[b]) then parent[b] = a;
4.     else{
5.       parent[a] = b;
6.       if (r[a] == r[b]) then r[b] = r[b] + 1;
7.     }
8. }
    
```

Union by rank to reduce the height of tree

### Disjoint set

```

1. union(a, b) { //merging 2 sets: the set with id=a and the set with id=b
2.   if (a != b) then
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```

Union(0, 1)

Union(1, 2)

Union(2, 3)

### Disjoint set

```

1. union(a, b) //merging 2 sets: the set with id=a and the set with id=b
2.   if (a != b) then
3.     if (r[a] > r[b]) then parent[b] = a;
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7.     }
8. }

```

Union by rank to reduce the height of the tree

Union(0, 1)

Union(1, 2)

Union(1, 3)

### Kruskal: disjoint set

```

1. Kruskal(G=(V,E),w){
2.   E' = sort edges in ascending order of weight;
3.   E_t = {}
4.   for (u, v) in E' do {
5.     if E_t ∪ (u, v) does not contain cycle then
6.       E_t = E_t ∪ (u, v);
7.     if (|E_t| == |V| - 1) then break;
8.   }
9.   if (|E_t| < |V| - 1) then {
10.    print("ERROR: Graph G is not connected");
11.    return NULL;
12.   }
13.   else return (V, E_t);
14. }

```

Not use Disjoint set

$O(|E| \log |E| + |E|(|E| + |V|))$

```

1. Kruskal(G=(V,E),w){
2.   for (x = 1; x <= |V|; x++) do makeSet(x);
3.   E' = sort edges in ascending order of weight;
4.   E_t = {}
5.   for (u, v) in E' do {
6.     idu = find(u);
7.     idv = find(v);
8.     if (idu != idv) then {
9.       union(idu, idv);
10.      E_t = E_t ∪ (u, v);
11.      if (|E_t| == |V| - 1) then break;
12.    }
13.   }
14.   if (|E_t| < |V| - 1) then {
15.    print("ERROR: Graph G is not connected");
16.    return NULL;
17.   }
18.   else return (V, E_t);
19. }

```

Use Disjoint set

What is the computation time if disjoint set is used ?

$O(|V| + |E| \log |E| + |E| |V|)$

$O(|V| + |E| \log |E| + |E| \log |V|)$

$O(|V| + |E| \log |E| + |E| \alpha(|V|))$

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### Kruskal: disjoint set

```

1. makeSet(x){ //Create a set containing only element x
2.   parent[x] = x;
3. }
4. find(x){ //not compression
5.   while (x != parent[x]) do x = parent[x];
6.   return x;
7. }
8. union(a, b) //not union by rank
11.   if (a != b) then
12.     parent[a] = b;

```

not optimize

$O(|V| + |E| \log |E| + |E| |V|)$

```

1. Kruskal(G=(V,E),w){
2.   for (x = 1; x <= |V|; x++) do makeSet(x);
3.   E' = sort edges in ascending order of weight;
4.   E_t = {}
5.   for (u, v) in E' do {
6.     idu = find(u);
7.     idv = find(v);
8.     if (idu != idv) then
9.       union(idu, idv);
10.      E_t = E_t ∪ (u, v);
11.      if (|E_t| == |V| - 1) then break;
12.    }
13.   }
14.   if (|E_t| < |V| - 1) then {
15.    print("ERROR: Graph G is not connected");
16.    return NULL;
17.   }
18.   else return (V, E_t);
19. }

```

If use union(a, b) not by rank and find(x) not compression:

- find(x): could take  $O(|V|)$  in a worst-case scenario (a long, skinny tree).
- union(a, b): Each union(a, b) operation takes  $O(1)$  because we just update the parent.

→ Kruskal: We call find(x) multiple times → the loop lines 5-13: the overall cost can be  $O(|E| |V|)$ .

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### Kruskal: disjoint set

```

1. makeSet(x){ //Create a set containing only element x
2.   parent[x] = x; r[x] = 0;
3. }
4. find(x){ //not compression
5.   while (x != parent[x]) do x = parent[x];
6.   return x;
7. }
8. union(a, b) //union by rank
11.   if (a != b) then
12.     if (r[a] > r[b]) then parent[b] = a;
13.     else{
14.       parent[a] = b;
15.       if (r[a] == r[b]) then r[b] = r[b] + 1;
16.     }
17. }

```

If use union(a, b) by rank and find(x) not compression:

- find(x):  $O(\log |V|)$
- union(a, b):  $O(1)$  to update parent and rank.

$O(|V| + |E| \log |E| + |E| \log |V|)$

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## Kruskal: disjoint set

```

makeSet(x) { //Create a set containing only element x
    parent[x] = x; r[x] = 0;
}

find(x) { //path compression
    if (x != parent[x])
        parent[x] = find(parent[x])
    return parent[x];
}

union(a, b) { //union by rank
    if (a != b) then
        if (r[a] > r[b]) then parent[b] = a;
        else{
            parent[a] = b;
            if (r[a] == r[b]) then r[b] = r[b] + 1;
        }
    }
}
    
```

```

1. Kruskal(G=(V,E),w){
2.   for (x = 1; x <= |V|; x++) do makeSet(x);
3.   E' = sort edges in ascending order of weight;
4.   Es = ∅
5.   for (u, v) in E' do {
6.       idu = find(u);
7.       idv = find(v);
8.       if (idu != idv) then {
9.           union (idu, idv);
10.          Es = Es ∪ {u,v};
11.          if (|Es| == |V| - 1) then break;
12.      }
13.  }
14.  if (|Es| < |V| - 1) then {
15.      print("ERROR: Graph G is not connected");
16.      return NULL;
17.  }
18.  else return (V, Es);
19. }
    
```

$$O(|V| + |E|\log|E| + |E| \alpha(|V|))$$

If use union(x, y) by rank and find(x) path compression:

- find(x):  $\alpha(|V|)$  where  $\alpha(|V|)$  is the **inverse Ackermann function** (If  $|V| \approx 10^{80}$  then  $\alpha(|V|) \leq 5$ ), so find(x) takes almost constant  $O(1)$ .
- union(a, b):  $O(1)$