

```
def find_cycle(G):
    #make set of all!
    #for each edge, call find_set(v), find_set(u)
    # if find_set(v) < findset(u), has cycle
    # else union(u, v)
```

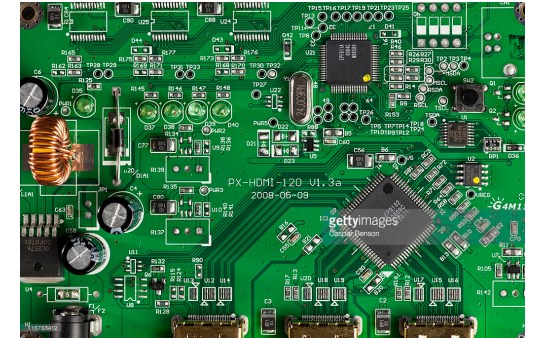
```
def find_cycle(G):
    vertices =
    for i in range(0, vertex):
        make_set(i)
    for i in range(0, vertex):
        for j in range (0, vertex):
            if a[i, j] =
```

Minimum Spanning Tree Overview

A Tree Spanning All Vertices Of A Given Connected Graph With Minimum Total Weight

Minimum Spanning Tree (MST) Overview

- A motivating example: To interconnect a set of n pins in an electronic circuit.



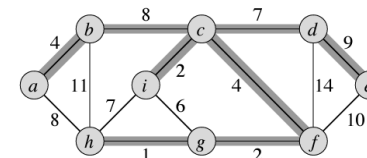
Minimum Spanning Tree (MST) Overview

- A motivating example: To interconnect a set of n pins in an electronic circuit
- Given a connected, undirected graph $G = (V, E)$ and a weight $w(u, v)$ for each edge $(u, v) \in E$, specifying the cost to connect (or traverse) u and v ,
- We want to find an acyclic subset of edges $T \subseteq E$ that connects all vertices in V and whose total weight $w(T)$ is minimized:

$$w(T) = \sum_{(u,v) \in T} w(u, v)$$

- Since T is acyclic and connects all vertices, it must form a tree, which we call a **spanning tree**, and a **minimum(-weight) spanning tree**.

MST Example (CLRS Fig. 23.1)



- Total weight (minimum) is 37. Shaded edges are tree edges.
- The MST is not necessarily unique
 - Remove (b, c) and add (a, h) .
 - You still get a spanning tree, and its edge-weight total is also 37.

Strategy To Find an MST

- We actually grow an MST. Of course it's not an MST until it's fully grown. It's just a subgraph of the final MST we'll get.
- Starting from the smallest subset of edges $A = \emptyset \subseteq T \subseteq E$,
 - We grow A (the subgraph that's being grown) to T (a full MST)
 - By adding one edge to A at a time (at each iteration of a loop).
- Surprisingly, there's a greedy choice property that makes this strategy possible.
 - That is, given a subset (of edges) A of an MST, we can always find an edge (u, v) such that $A \cup \{(u, v)\}$ is still a subset of an MST.
 - This edge (u, v) is called a **safe edge** for A .
 - Since we can add it safely to A while maintaining the invariant (A being a subset of an MST)

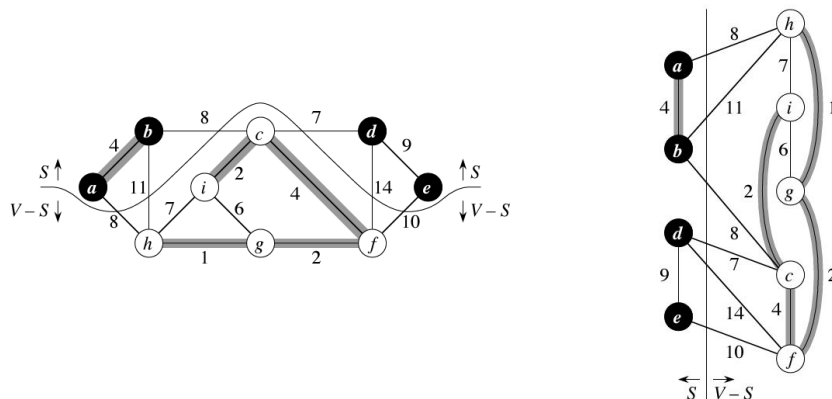
Generic MST Finding Logic

```

GENERIC-MST( $G, w$ )      //  $G = (V, E), w: E \rightarrow \mathbb{R}$  (edge-weight function)
1   $A = \emptyset$ 
2  while  $A$  does not form a spanning tree      // i.e., while  $|A| < |G.V| - 1$ 
3      find an edge  $(u, v)$  that is safe for  $A$ 
4       $A = A \cup \{(u, v)\}$ 
5  return  $A$ 
    
```

- Of course line 3 is not trivial.
- Theorem 23.1 gives us a general strategy to find a safe edge.
 - And it's a greedy choice (thus all MST algorithms we will learn are greedy)
 - Need to understand the following:
 - A cut of vertices
 - An edge crossing a cut
 - A set of edges respecting a cut
 - A light edge crossing a cut.

Cut, Cut-Crossing Edge, Mutually Respecting Cut & Edge Set, Light Cut-Crossing Edge (CLRS Fig. 23.2)

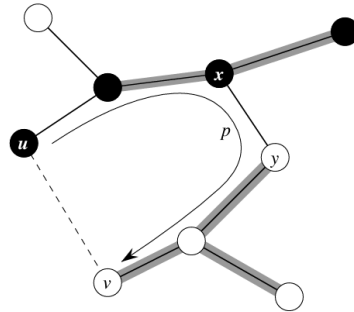


Greedy Choice Property of MST Algorithms (Theorem 23.1)

- Given a cut $(S, V - S)$ and a cut-respecting edge set A which is also a subgraph of an MST, a light cut-crossing edge (u, v) is safe!
- A lot of terminology, but bottom line is:
 - Given a subgraph A of an MST, **find a cut (any cut) that respects A** (no edges in A crossing that cut). Let's call the cut $(S, V - S)$.
 - Find all edges crossing the cut. Pick a light edge of those edges (minimum weight). Call such a light edge (u, v) .
 - Then $A \cup \{(u, v)\}$ is still a subgraph of an MST.
 - Repeat this $|V| - 1$ times from $A = \emptyset$, and you get an MST!
- It's a greedy strategy (picking a minimum-weight crossing edge).

Proof Sketch (CLRS Fig. 23.3)

- “Cut-and-paste” technique.
- Assume T is an MST which A is a subgraph of, but not including a light edge (u, v) for a cut $(S, V - S)$.
- Then we can “cut” the edge in the MST T crossing the cut. Remove (“cut”) that edge, and add (“paste”) the light edge (u, v) . Call the resulting tree T' .
- Then we can show that T' is still an MST.



Multiple MST Algorithms Possible

- Because there are many ways to **form A and find a cut (any cut) that respects A !**
- Kruskal’s strategy:
 - Given a **sub-forest** of an MST, find all edges that connect two trees, pick a minimum-weight edge and add it to the forest.
 - Starting from $|V|$ singleton trees, reduce # trees by 1 at every iteration, ending with only 1 spanning tree, which must be an MST (Theorem 23.1).
 - The cut here is a partition of the forest of all trees, which the minimum-weight edge crosses.

Sub-Forest (Kruskal) vs. Sub-Tree (Prim)

- Prim’s strategy:
 - Given a **sub-tree** of an MST, find all edges that connect a vertex in the tree to a non-tree vertex, pick a minimum-weight edge of all and add it to the tree.
 - Starting from one singleton tree (root), grow the tree’s size by 1 at every iteration, ending with 1 spanning tree, which must be an MST (Theorem 23.1).
 - The cut here is the growing sub-tree of an MST, and rest of vertices.

Kruskal’s MST Algorithm

Add Minimum-Weight Edge To Ongoing MST Subgraph (Forest) If It Doesn’t Form A Cycle. Discard It If It Forms A Cycle. Then Proceed To Next Smallest-Weight Edge.

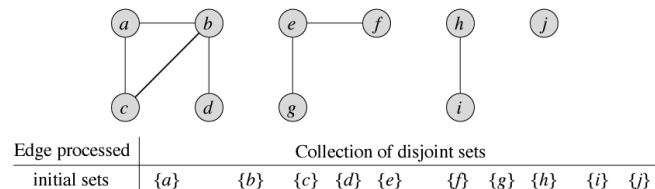
Strategy of Kruskal's MST Algorithm

- Finding a cut that respects the ongoing MST subgraph is not really important.
- Focus on a light edge. What's the first candidate?
 - Minimum-weight edge of all remaining.
- If a minimum-weight edge doesn't form a cycle, it must be a safe (because it doesn't form a cycle) and light (because it's minimum-weight) edge!
- If it forms a cycle, throw it away (can't use it anyway) and try the next minimum-weight edge.
 - It would be handy to sort edges in non-decreasing order of weights, then scan them one by one.
- The key point here is **how to check if adding an edge forms a cycle**.

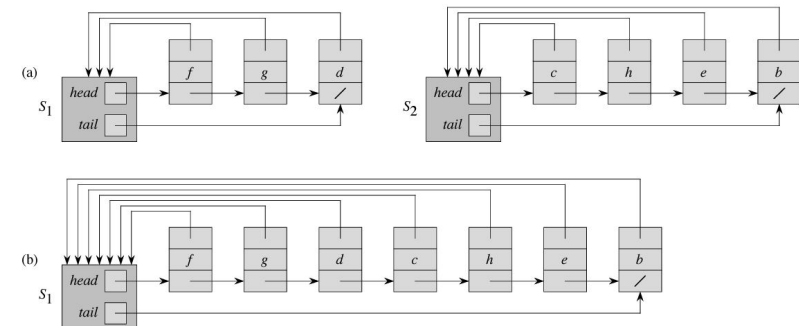
Disjoint-Set Data Structure (CLRS Ch. 21)

- Efficiently maintains disjoint sets of elements (vertices).
- Supports three operations:
 - MAKE-SET(x): Make a singleton set with one element x . Trivial ($O(1)$).
 - FIND-SET(x): Returns a representative (unique) element from the set that contains x . Time complexity depends on how to implement the data structure.
 - UNION(x, y): Returns the union of two **disjoint** sets S_x and S_y (where $x \in S_x$ and $y \in S_y$). Time complexity depends on how to implement.
- Then, the idea is to form disjoint sets of vertices each of which corresponds to a tree in the ongoing forest in MST algorithm.
 - Then checking if adding (u, v) to the forest would form a cycle is equivalent to check whether $\text{FIND-SET}(u) = \text{FIND-SET}(v)$!

Disjoint-Set Concept & Example: Connected Components (CLRS Fig. 21.1)

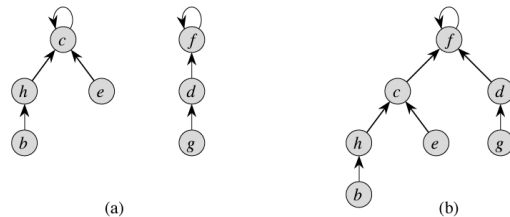


Linked-List Representation Of Disjoint Sets (CLRS Ch. 21-2, Fig. 21.2)



- FIND-SET(x) is $O(1)$, UNION(x, y) is $O(\max(|S_x|, |S_y|))$.

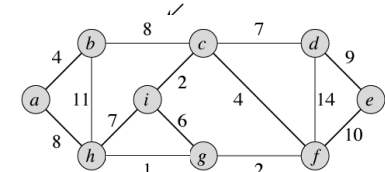
Disjoint-Set Forests (CLRS Ch. 21-3, Fig. 21.4)



- **FIND-SET**(x) is $O(h)$ (where h is height of the tree).
- **UNION**(x, y) is $O(h)$ for the above straightforward idea.
- But this is NOT an improvement over linked-list representation.
- There are heuristics to improve running time (study CLRS Ch. 21-3).
 - Giving $O(m\alpha(n))$, where $\alpha(n)$ is a very slowly growing function (Ch. 21-4: optional)
 - m : sum of # MAKE-SET, UNION, and FIND-SET ops. n : # MAKE-SET ops.

Kruskal's Algorithm Example (CLRS Fig. 23.4)

Sorted edges: $(h, g): 1, (c, i): 2, (g, f): 2, (a, b): 4, (c, f): 4, (g, i): 6, (c, d): 7, (h, i): 7, (a, h): 8, (b, c): 8, (d, e): 9, (e, f): 10, (b, h): 11, (d, f): 14$



MST-KRUSKAL(G, w)

```

1   $A = \emptyset$ 
2  for each vertex  $v \in G.V$ 
3      MAKE-SET( $v$ )
4  sort the edges of  $G.E$  into nondecreasing order by weight  $w$ 
5  for each edge  $(u, v) \in G.E$ , taken in nondecreasing order by weight
6      if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ )
7           $A = A \cup \{(u, v)\}$ 
8          UNION( $u, v$ )
9  return  $A$ 
```

Time Complexity of Kruskal's Algorithm

- $O(E)$ **FIND-SET** and **UNION** operations in for loop (line 5-8).
- $|E| \geq |V| - 1$, because G is assumed to be connected.
- Total $O(E\alpha(V))$ (disjoint-set forest representation with heuristics).
- Since $\alpha(V) = O(\lg V)$, total is $O(E \lg V)$.

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```

Prim's MST Algorithm

Ongoing MST Subgraph Is Always a Tree. Add To The Tree a Minimum-Weight Edge that Will Still Form a Tree.

Strategy Of Prim's MST Algorithm

- The ongoing MST subgraph always forms a regular tree (MST subtree).
- Then at each iteration, we add to the tree a new *connected* edge which will still form a tree (no cycle).
- Since the newly added edge should be safe (i.e., the resulting bigger tree should still be a subgraph of an MST), it must be a minimum-weight one of all possible such edges.
- Starting from a singleton MST subtree (root vertex only), add one edge at a time until the tree includes all vertices.
- The key point here is how to find such an edge.

Maintaining Not-Yet-Included Vertices In The Order Of Proximity To Current MST Subtree

- Observation: When a new vertex is added (moved) to the ongoing/growing MST subtree, only its adjacent vertices might get new proximity values (minimum weight to any tree vertex).
- The not-yet-included vertices can be maintained in a min-priority queue (heap).
- When the minimum is extracted from the heap, adjust the proximity values of all its (the minimum's) adjacent vertices in the heap.

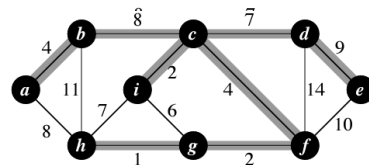
Prim's MST Algorithm And Example

$Q =$

MST-PRIM(G, w, r)

```

1  for each  $u \in G.V$ 
2     $u.key = \infty$ 
3     $u.\pi = \text{NIL}$ 
4   $r.key = 0$ 
5   $Q = G.V$ 
6  while  $Q \neq \emptyset$ 
7     $u = \text{EXTRACT-MIN}(Q)$ 
8    for each  $v \in G.Adj[u]$ 
9      if  $v \in Q$  and  $w(u, v) < v.key$ 
10          $v.\pi = u$ 
11          $v.key = w(u, v)$  // This assignment may cause rearrangement of vertices in  $Q$ .
```



Time Complexity Of Prim's Algorithm

- Depends on how to implement the min-priority queue Q .
- If we use a binary min-heap (CLRS Ch. 6),
 - Line 1-5: BUILD-MIN-HEAP for $O(V)$ time.
 - Line 6 while loop iterates $|V|$ times. Line 7 EXTRACT-MIN for $O(\lg V)$.
 - Giving $O(V \lg V)$ for EXTRACT-MIN.
 - Line 8 for loop iterates $O(E)$ times all together (amortized): $\sum |Adj(v)| = 2|E|$
 - Line 9 (Q membership test) can be $O(1)$ (maintain flags)
 - Line 11: DECREASE-KEY on the min-heap: $O(\lg V)$
 - $\therefore O(V \lg V + E \lg V) = O(E \lg V)$.

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```

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