

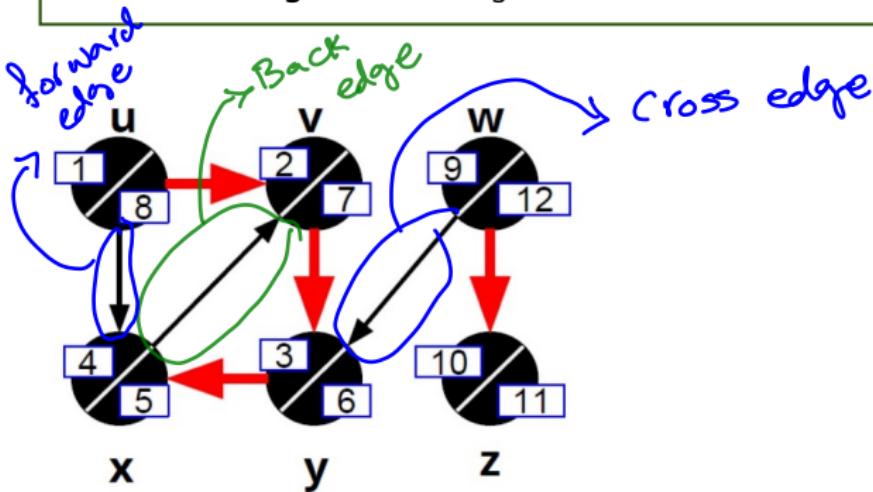
CSC263H

Data Structures and Analysis

Prof. Bahar Aameri & Prof. Marsha Chechik

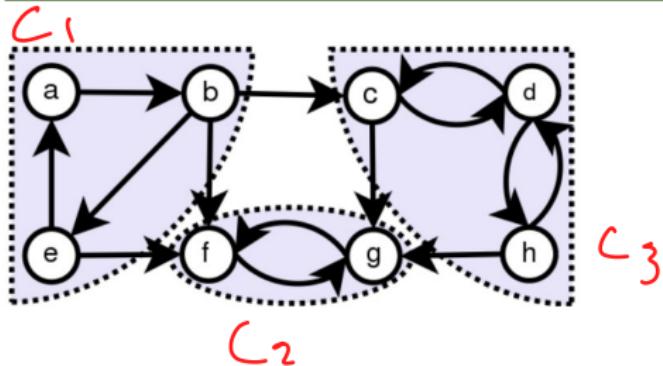
Winter 2024 – Week 10

- **Tree edge:** an edge in the DFS-forest
- **Back edge:** a **non-tree** edge pointing from a vertex to its **ancestor** in the DFS forest.
- **Forward edge:** a **non-tree** edge pointing from a vertex to its **descendant** in the DFS forest.
- **Cross edge:** all other edges.



Finding Strongly Connected Components

A **strongly connected component** of a directed graph $G = \langle V, E \rangle$ is a maximal set of vertices C , such that for every pair of vertices u and v , there's path from u to v and a path from v to u , i.e. vertices v and u are reachable from each other.



$$C_1 = \{a, b, e\}$$

$$C_2 = \{f, g\}$$

Finding Strongly Connected Components

- The transpose of a directed graph $G = (V, E)$ is the graph $G^{\text{reverse}} = (V, E^{\text{reverse}})$, where

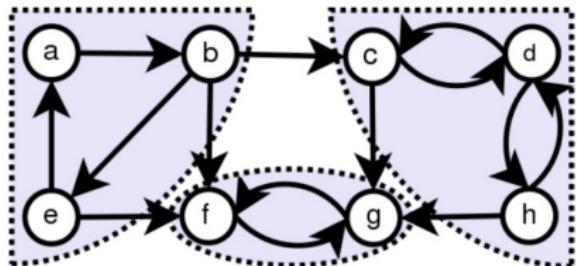
$$G^T$$

$$E^{\text{reverse}} = \{(v, u) \in V \times V : (u, v) \in E\}$$

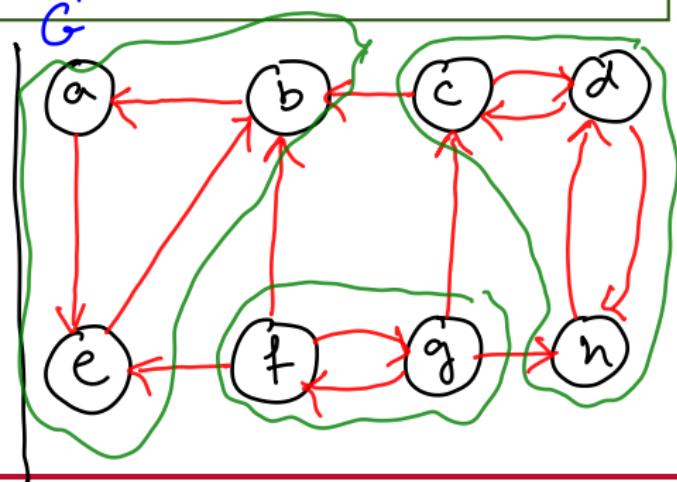
That is, G^{reverse} is G with all edges reversed.

- A directed graph G and its transpose G^{reverse} have the same Strongly Connected Components.

G



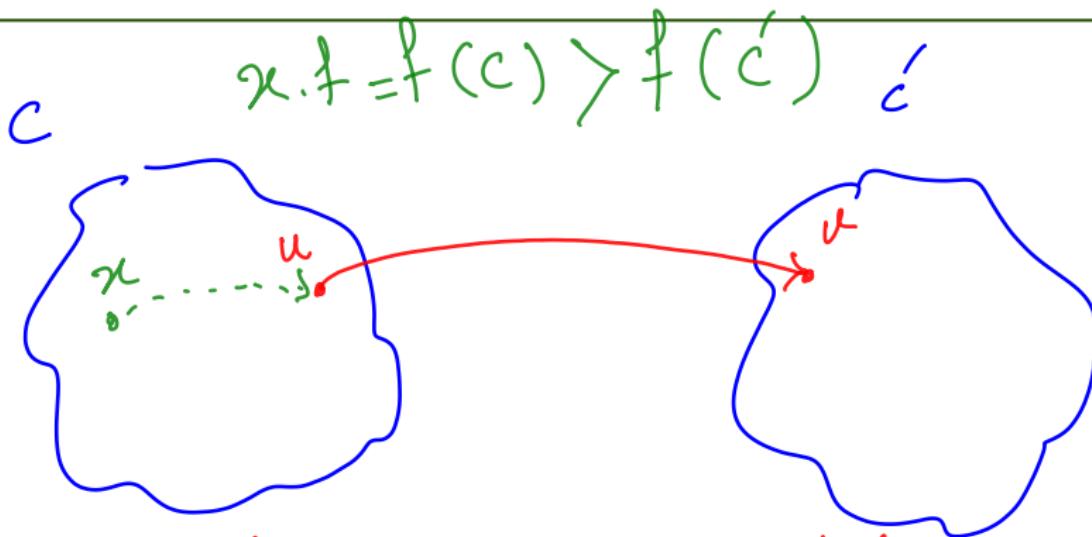
G^T



Finding Strongly Connected Components

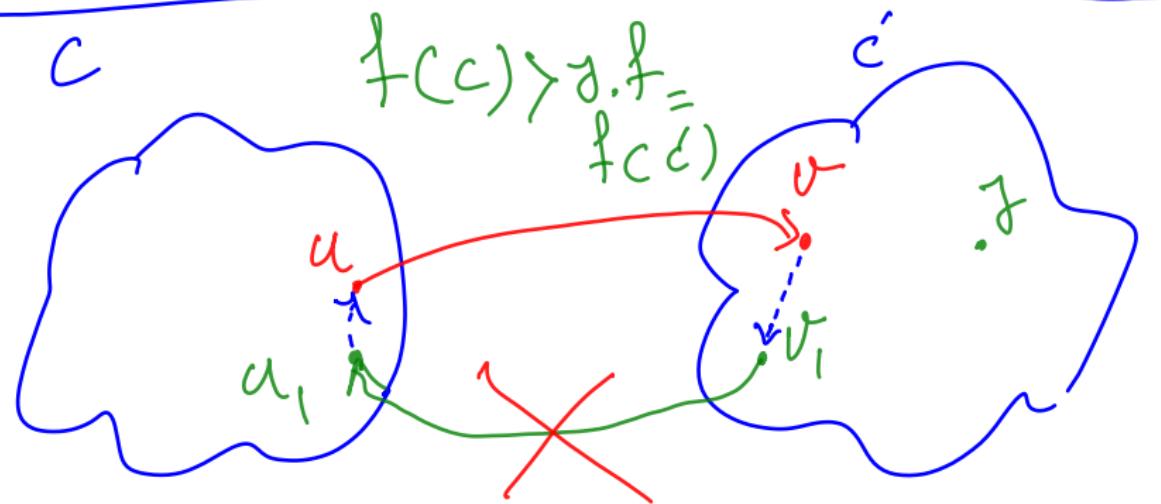
Let C and C' be distinct SCC's in $G = \langle V, E \rangle$. Suppose there is an edge $(u, v) \in E$ such that $u \in C$ and $v \in C'$.

Then $f(C) > f(C')$, where $f(U) = \max_{u \in U} \{u.f\}$
(latest finishing time of any vertex in U).



Case 1: DFS discovers a vertex in C before discovering any vertex in C'

Case 2: $N \sim \sim \sim \sim \sim C'$
 $\sim \sim \sim \sim \sim C$

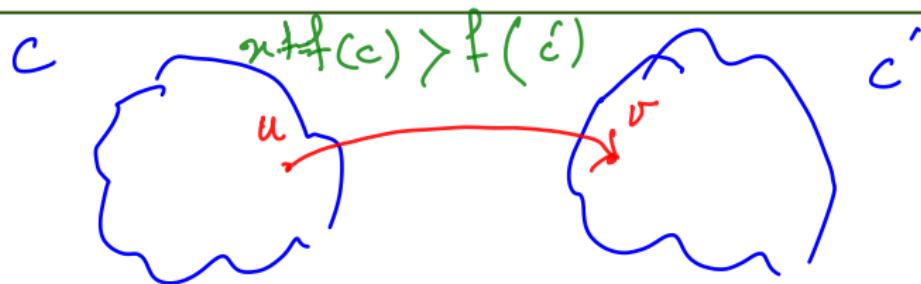


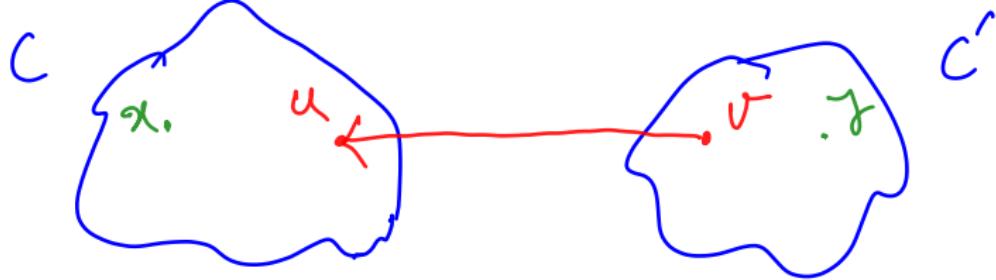
Complete Proof in CLRS

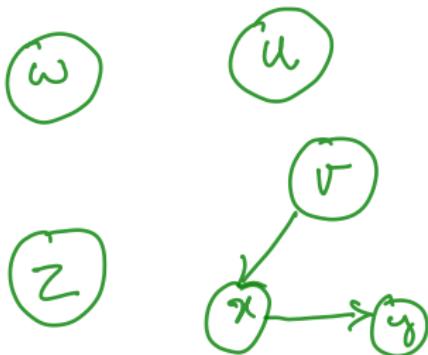
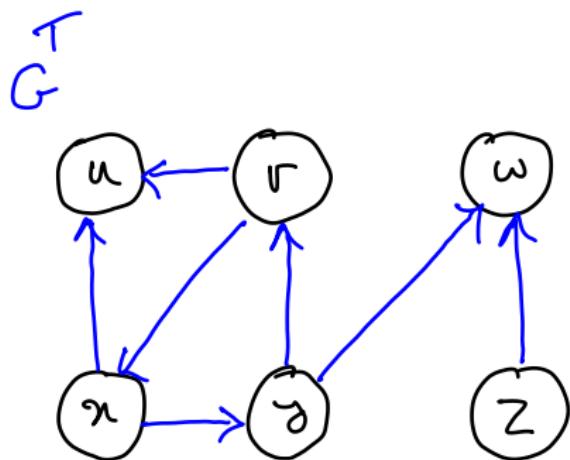
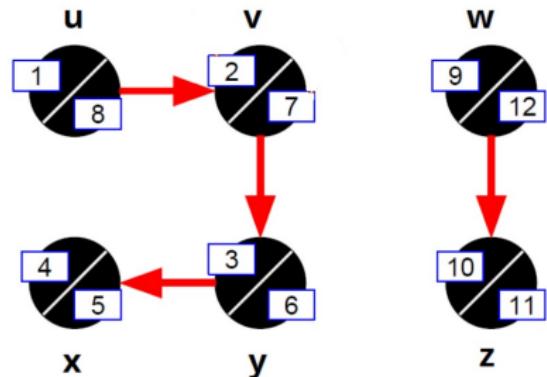
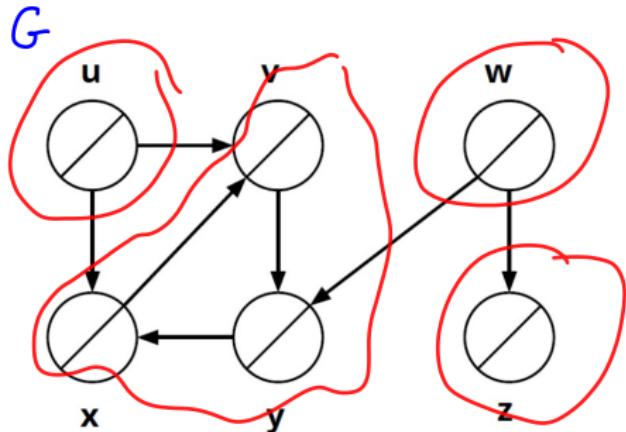
Finding Strongly Connected Components

StronglyConnectedComponents(G)

1. Call $DFS(G)$ to compute finishing times $u.f$ for all u .
2. Compute $G^{reverse}$
3. Call $DFS(G^{reverse})$, but in the main loop, consider vertices in order of decreasing $u.f$ (as computed in the first DFS)
4. Output the vertices in each tree of the depth-first forest formed in second DFS as a separate SCC.

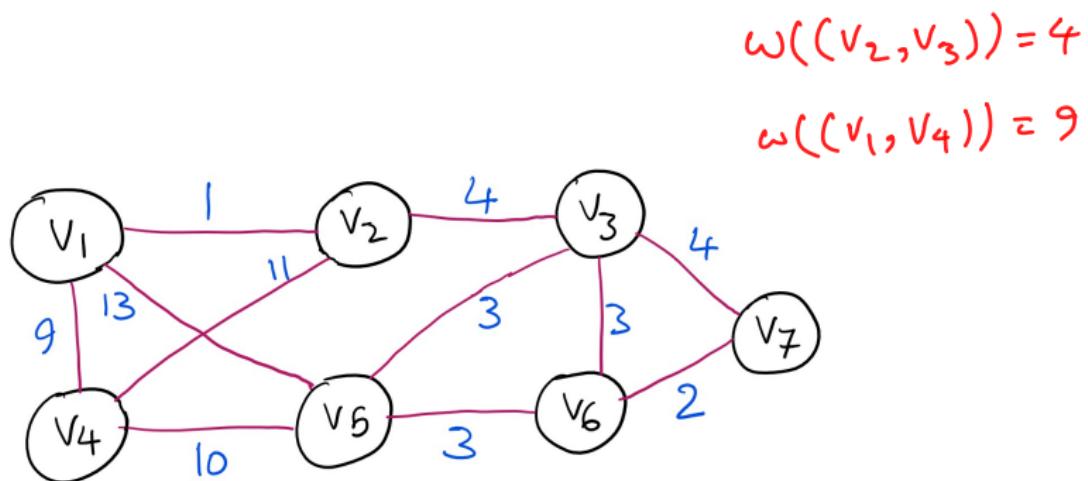






Review: Weighted Graphs

A **weighted graph** is a graph $G = \langle V, E \rangle$ for which each edge has an associated **weight**, given by a weight function $w : E \rightarrow \mathbb{R}$.



Minimum Spanning Trees (MSTs)

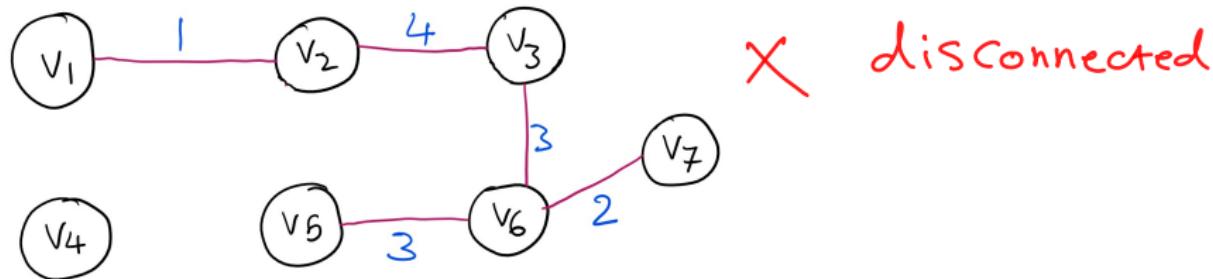
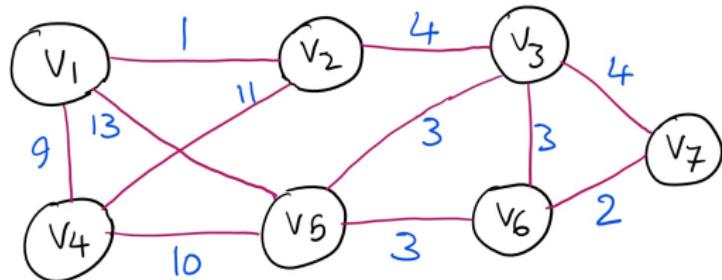
Let $G = \langle V, E \rangle$ be a **connected undirected weighted** graph.

A **minimum spanning tree** T for G is a sub-graph of G which

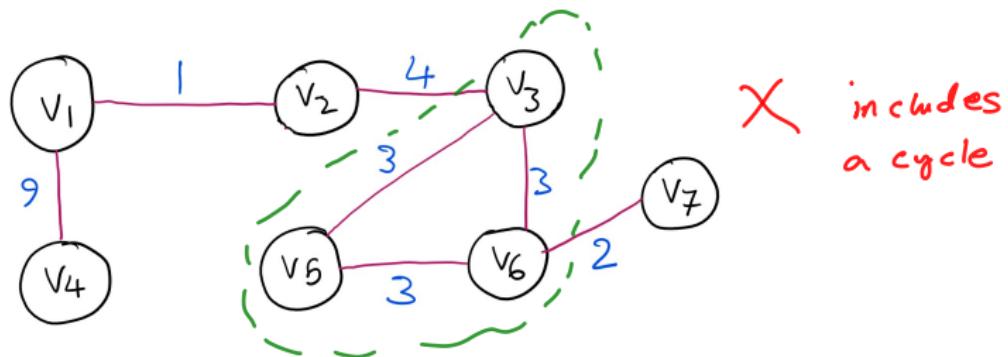
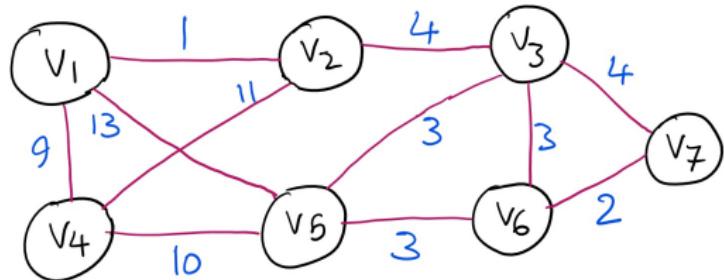
- includes all vertices of G ; \longrightarrow **Spanning**
- is acyclic and connected; \longrightarrow **Tree**
- its total weight is minimized. \longrightarrow **Minimum**
(among all other sub-graph of G with the above two properties).

Note: A graph might have more than one MST.

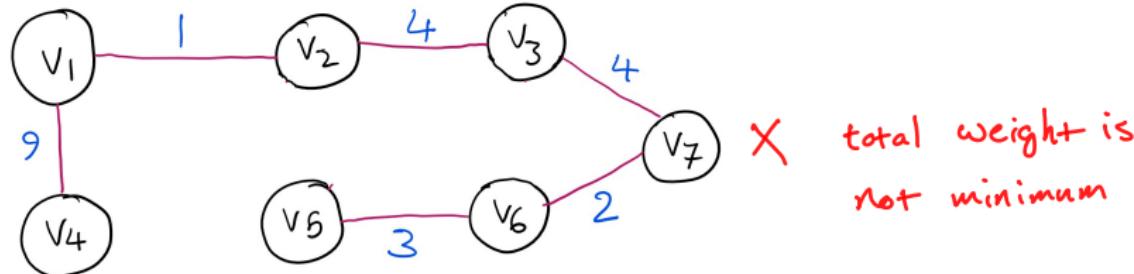
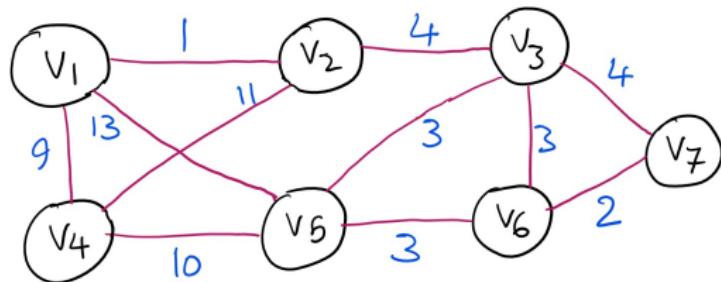
Minimum Spanning Trees (MSTs): Examples



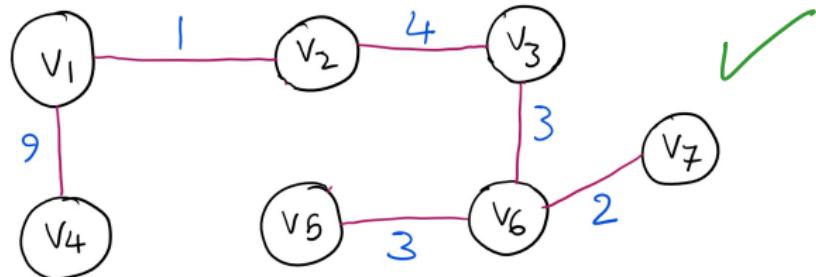
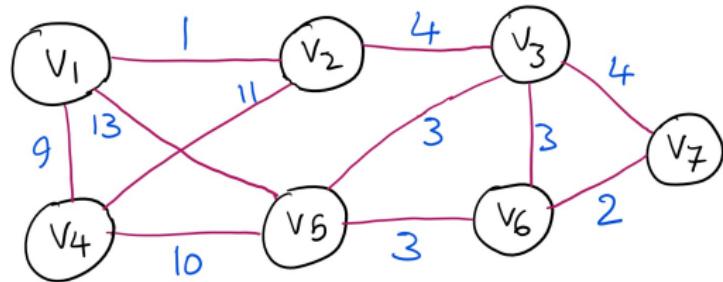
Minimum Spanning Trees (MSTs): Examples



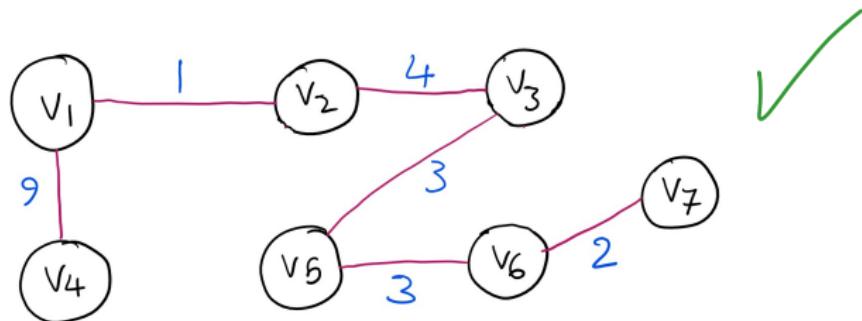
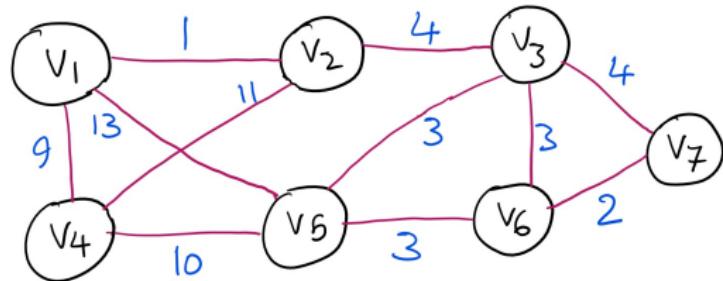
Minimum Spanning Trees (MSTs): Examples



Minimum Spanning Trees (MSTs): Examples



Minimum Spanning Trees (MSTs): Examples



Minimum Spanning Trees: Some Applications

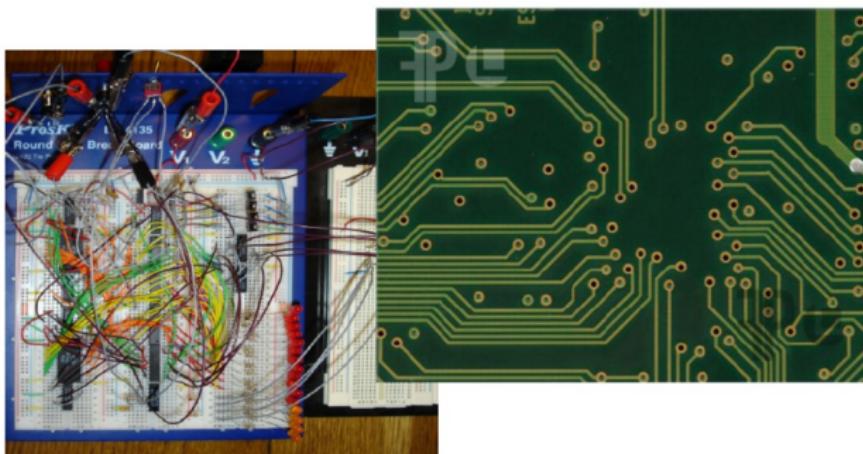
Finding Minimum Costs/Distances

- **Example 1:** Build a road network that connects all towns and with the minimum cost.



Minimum Spanning Trees: Some Applications

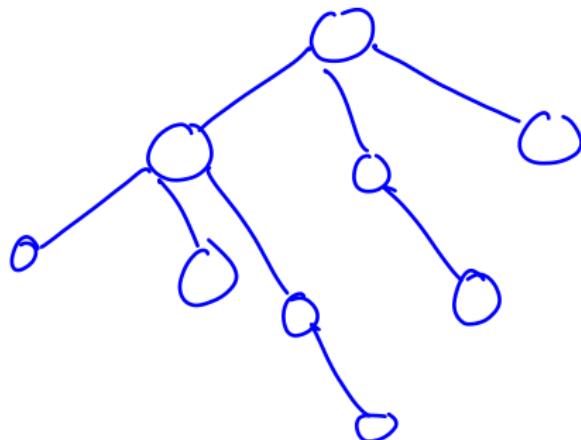
- **Example 2:** Connect all components of a circuit with the least amount of wiring.



Review: Properties of Trees

Let T be a tree with n vertices. Then

- T has exactly $n-1$ edges.
- Adding one edge to T will create a cycle
- Removing one edge from T will disconnect the graph



- The MST of a connected undirected graph $G = \langle V, E \rangle$ has $|V|$ vertices.
- The MST of a connected undirected graph $G = \langle V, E \rangle$ has $|V| - 1$ edges.

Finding MSTs of a Graph: Ideas

Idea 1: Let $T.V = G.V$. Start with $T.E = G.E$; Keep **deleting** edges until an MST remains.

$$|V| = n$$

In the **worst-case**, how many **edges** are deleted?

$$\frac{n(n-1)}{2} - (n-1) \in \Theta(n^2)$$

An undirected graph G with n vertices has at most $\frac{n(n-1)}{2}$ edges.

Idea 2: Let $T.V = G.V$. Start with **empty** $T.E$; Keep **adding** edges until an MST is built.

In the **worst-case**, how many **edges** is added?

$$n-1 \in \Theta(n)$$

GenericMST(G):

1. $T.V = G.V$ **# Initializing vertices of the MST**
2. $T.E = \emptyset$ **# Starting with an empty set of edges**
3. **While** T does not form a spanning tree:
4. find an edge $e \in G.E$ that is **safe** for T
5. $T.E = T.E \cup \{e\}$ **# Add the edge e to the MST**
6. **return** T

Challenge: What is a **safe** edge? and how to **find** one?

Growing MSTs: Finding Safe Edges

Intuition: While growing T , if we make sure that T is always a **subgraph** of some MST of G , then eventually T will become an MST.

Analogy: If we make sure the pieces we put together is always a subset of the real picture while we grow it, then eventually it will become the real picture! (Picture borrow from Larry Zhang's Slides)



Edge e is **safe** for T iff $T.E \cup \{e\}$ is a subset of edges of some MST of G .

Growing MSTs: Finding Safe Edges

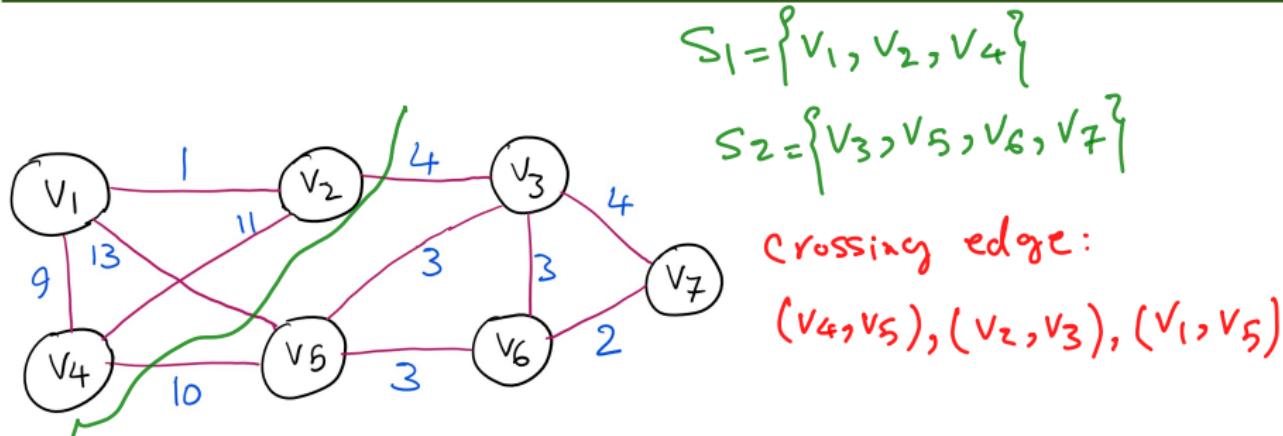
A **cut** $\langle S_1, S_2 \rangle$ of an undirected graph $G = \langle V, E \rangle$ is a partition of V into two **disjoint** subsets S_1 and S_2 .

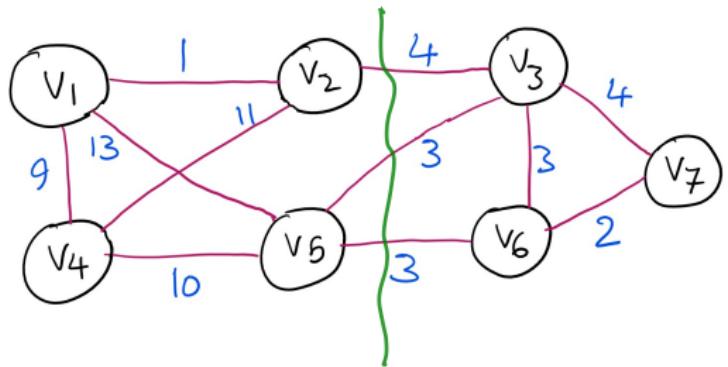
That is, $S_1 \cap S_2 = \emptyset$ and $S_1 \cup S_2 = V$.

We can also define S_2 in terms of V and S_1 . That is, $S_2 = (V - S_1)$.

Example: If a graph G has exactly two **connected components**, then the two components form a cut of G .

An edge $(u, v) \in E$ **crosses** the **cut** $\langle S_1, S_2 \rangle$ if one of its endpoints is in S_1 and the other is in S_2 .





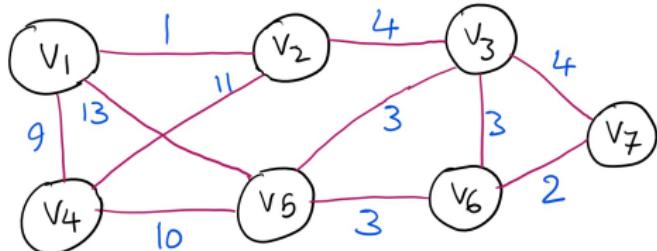
Theorem: Let G be a connected undirected weighted graph, and T be a subgraph of G which is a subset of some MST of G .

Let edge e be the **minimum weighted** edge among all edges that **cross** different **connected components** of T .

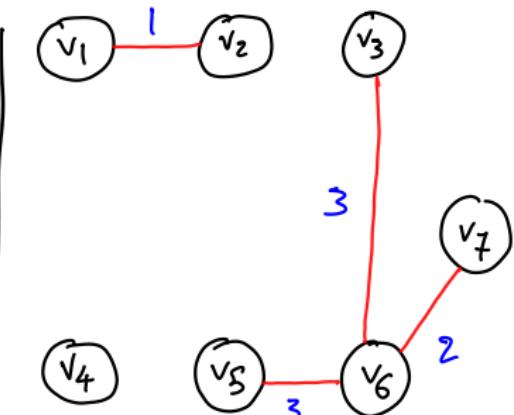
Then e is **safe** for T .

Proof: Theorem 23.1 and Corollary 23.2 in CLRS.

$G:$

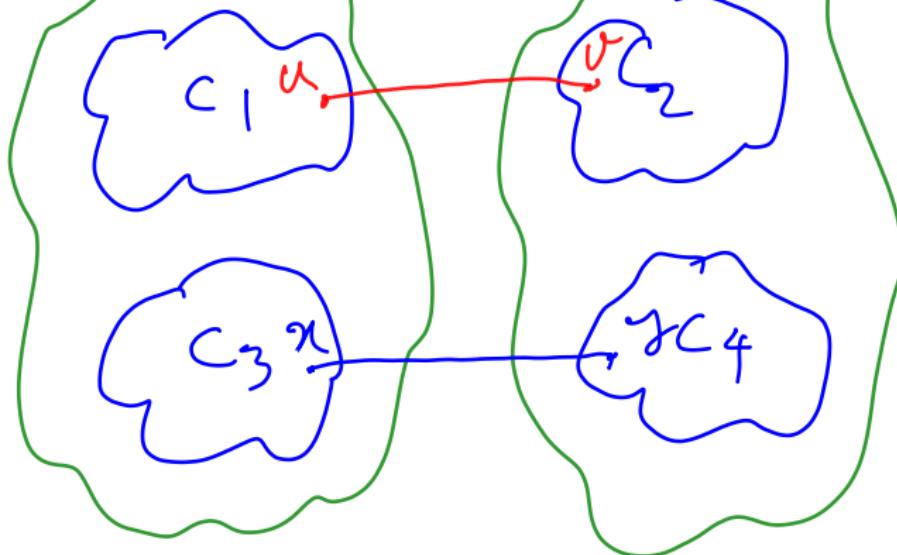


$T:$



Safe edge: (v_2, v_3)

T_{S_1}



S_2

T_{mst}

$$(x, y) \in T_{mst}$$

$$(u, v) \notin T_{mst}$$

$$(x, y) \notin T'_{mst}$$

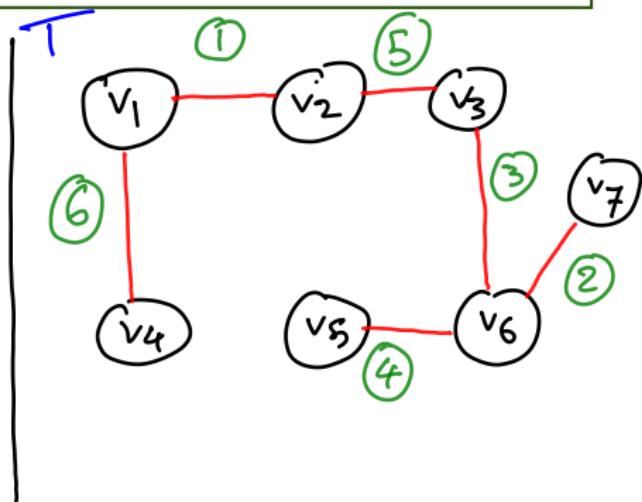
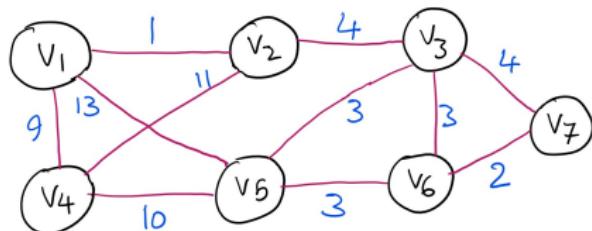
$$(u, v) \in T'_{mst}$$

The Generic MST Growing Algorithm

GenericMST(G):

1. $T.V = G.V$ # Initializing vertices of the MST
2. $T.E = \emptyset$ # Starting with an empty set of edges
3. While T does not form a spanning tree:
4. find $e \in G.E$ with min weight that crosses two connected components of T
5. $T.E = T.E \cup \{e\}$ # Add the edge e to the MST
6. return T

G



For the actual implementation of the growing algorithm:

- How to keep track of the **connected components**?
- How to efficiently find the **minimum weighted** edge?

The (main) difference between [Kruskal's](#) algorithm and [Prim's](#) algorithm is that they use **different data structures** to do these two things.

Kruskal's MST Algorithm

High-Level Idea:

Go through the edges in order of non-decreasing **weight**;

Select and add each edge that **crosses** two **connected components** of T .

That is, if for the edge (u, v) , u is in **one** connected component and v is in **another**.

How to keep track of **connected components** of T efficiently?



ConnectedComponents(G):

1. **for** each vertex $v \in G.V$:
2. **MakeSet**(v)
3. **for** each edge $(u, v) \in G.E$:
4. **if** **FindSet**(u) \neq **FindSet**(v): # If u and v are in different connected components
5. **Union**(u, v)

Kruskal's MST Algorithm

High-Level Idea:

Go through the edges in order of non-decreasing **weight**;

Select and add each edge that **crosses** two **connected components** of T .

That is, if for the edge (u, v) , u is in **one** connected component and v is in **another**.

Implementation Idea:

1. **Sort** the edges according to their weights.
2. Go through the sorted list from **lightest** to **heaviest**:
 - Add (u, v) to T if $\text{FindSet}(u) \neq \text{FindSet}(v)$
 - **After** adding (u, v) to T , u and v belong to the **same** connected component of T ; Hence $\text{Union}(u, v)$

Pre-condition: G is an undirected graph.

KruskalMST(G):

1. $T.V = G.V$
2. $T.E = \emptyset$
3. **for** each vertex $v \in G.V$: # Create a separate connected component for each vertex
4. **MakeSet**(v)
5. Let e_1, e_2, \dots, e_m be the edges in $G.E$ sorted by **weight** in non-decreasing order
6. **for** $i = 1$ to m :
7. Let $(u, v) = e_i$
8. **if** $\text{FindSet}(u) \neq \text{FindSet}(v)$: # Check u and v are in different components of T
9. $T.E = T.E \cup \{(v, u)\}$ # Add the safe edge (v, u) to T
10. **Union**(u, v) # Now u and v are in the same connected component of T
11. **return** T

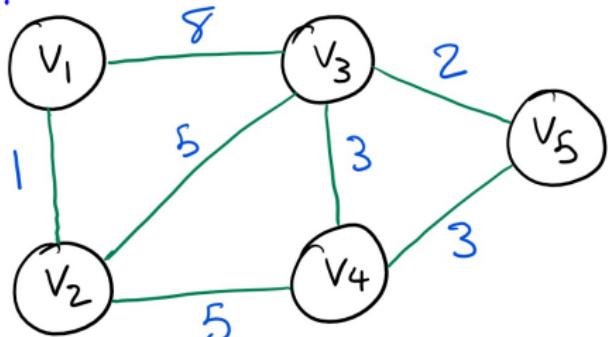
Len seq ↪ $K, r \rightarrow$ num of items in DS

$O(K \alpha(r))$ $K = 3m$ $O(3m \alpha(n))$

$r = n$ $O(m \alpha(n))$

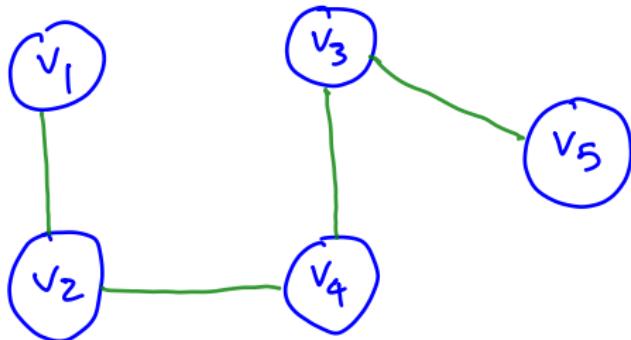
Edges sorted by weight: $(v_1, v_2), (v_3, v_5), (v_3, v_4), (v_4, v_5), (v_2, v_4), (v_2, v_3), (v_1, v_3)$

G:



Iteration	Edge
1	(v ₁ , v ₂)
2	(v ₃ , v ₅)
3	(v ₃ , v ₄)
4	(v ₄ , v ₅)
5	(v ₂ , v ₄)
6	(v ₂ , v ₃)
7	(v ₁ , v ₃)

T:

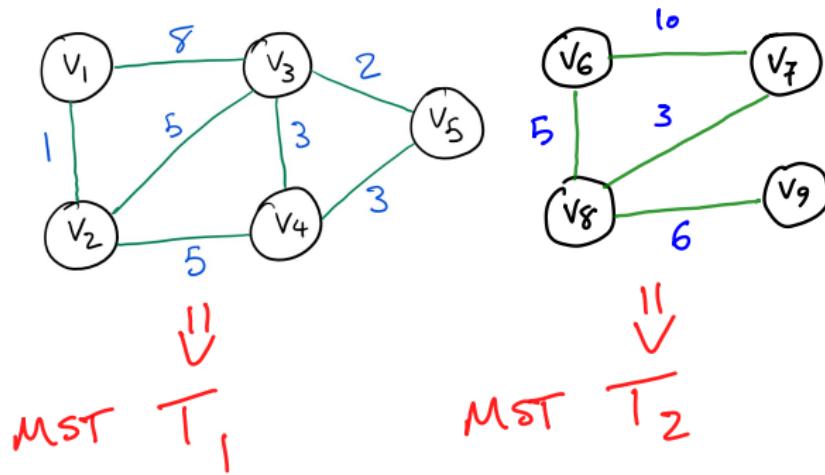


$\{v_1\}$ $\{v_2\}$ $\{v_3\}$ $\{v_4\}$ $\{v_5\}$
 $\{v_1, v_2\}$, $\{v_3, v_5, v_4\}$

Kruskal's MST Algorithm

What if the input graph G is **not connected**?

Kruskal's Algorithm will generate a **Minimum Spanning Forest**



Kruskal's MST Algorithm: Worst-case Run Time

Assuming that Disjoint-Sets implemented by tree with **union-by-rank** and **path-compression**.

1. Sorting edges: $\mathcal{O}(m \log m)$

$$|V|=n \quad |E|=m$$

2. **for** loop on Lines #3 and #4 (MakeSet): $\mathcal{O}(n)$

3. **for** loop on Lines #6 to #10 (FindSet and Union): $\mathcal{O}(m \alpha(n)) \simeq \mathcal{O}(m)$

Total running time: $\mathcal{O}(n + m \log m)$

Recall that $m \leq \frac{n(n-1)}{2}$. $m \in \mathcal{O}(n^2) \Rightarrow \log m \in \mathcal{O}(\lg n^2) \Rightarrow$
 $\lg n^2 = 2 \lg n \in \mathcal{O}(\lg n)$

So total run time can be **rewrite** as:

$\mathcal{O}(n + m \lg n)$

If the input graph G is **connect**, the run time is:

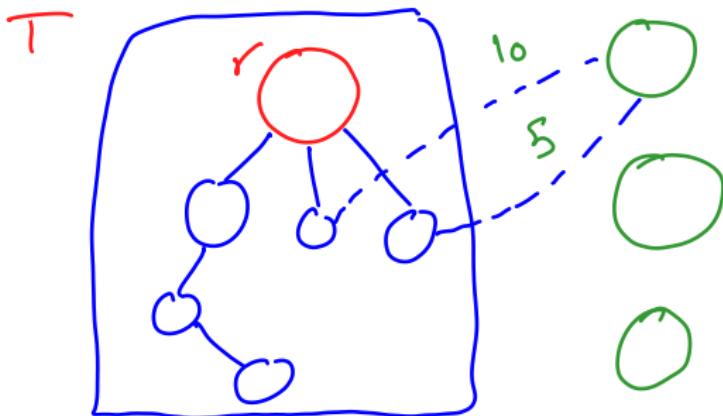
$m \geq n-1 \Rightarrow n \in \mathcal{O}(m) \quad \mathcal{O}(m \lg n)$

Prim's MST Algorithm

Pick an **arbitrary** vertex r of G as the starting vertex of T .

Grow T by adding one edge at a time:

- At every step T consists of one **connected component** that contains r , and **singleton** components that are **not** connected with r (each one is called an **isolated** vertex).
- At each step pick a **crossing** edge (between the component containing r and one of the **isolated** vertices) with the **minimum weight**.



Pick an **arbitrary** vertex r of G as the starting vertex of T .

Grow T by adding one edge at a time:

- At every step T consists of one **connected component** that contains r , and **singleton** components that are **not** connected with r (each one is called an **isolated** vertex).
- At each step pick a **crossing** edge (between the component containing r and one of the **isolated** vertices) with the **minimum weight**.

How to find the minimum crossing edge(s) efficiently?

Store all **isolated** vertices in a **min-heap**;

Keys of the min-heap are **weights** of the **crossing** edges (between isolated vertices and T).

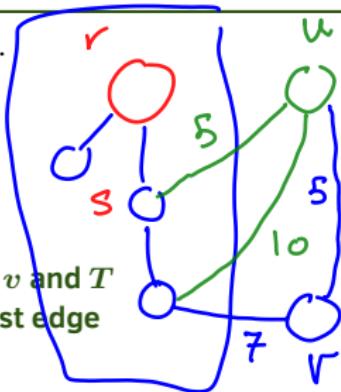
MinHeap Operations:

- BuildMinHeap(A): Given a list of elements A , return a **min-heap** that includes all elements in A . $\mathcal{O}(n)$
- ExtractMin(H): Remove and return the item from the heap H with the **lowest** key. $\mathcal{O}(\lg n)$
- DecreaseKey(H, i, k): **Decrease** the key value of the element x , stored in position i , to the new value k . (k assumed to be less than the current key of x). $\mathcal{O}(\lg n)$

Precondition: G is a **connected** undirected graph. r is a vertex of G .

PrimMST(G, r):

1. $T.V = G.V$
2. $T.E = \emptyset$
3. **for** each vertex $v \in G.V$:
4. $v.key = \infty$ # $v.key$ keeps the shortest distance between v and T
5. $v.p = \text{nil}$ # $v.p$ keeps who in T is v connected to via lightest edge
6. $r.key = 0$ # Set r to be the root of min-heap
7. $H = \text{BuildMinHeapGraph}(G)$
8. **While** H not empty:
9. $u = \text{ExtractMin}(H)$
10. **if** $u.p \neq \text{nil}$: # $u.p$ is nil only when $u = r$
11. $T.E = T.E \cup \{(u.p, u)\}$ # Add a safe edge to the MST
12. **for** each $v \in G.adj[u]$: # all u 's neighbours' distances to T need update
13. **if** $v \in H$ **and** $w((u, v)) < v.key$:
14. $v.p = u$
15. DecreaseKey($H, v.pos, w((u, v))$)
16. **return** T

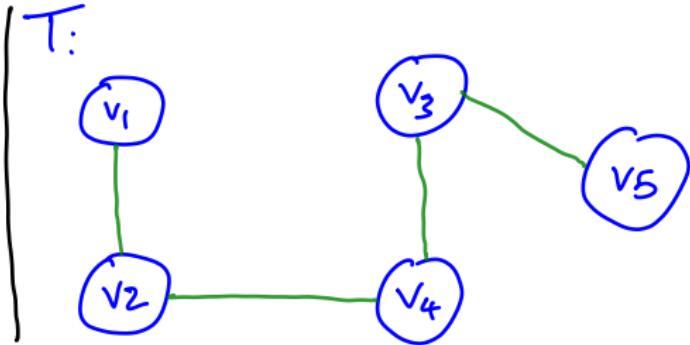
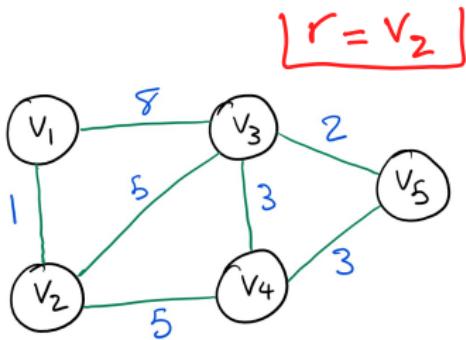


$$U.P = S$$

$$U.Key = 5$$

For each vertex v , access to v 's position in H can be achieved in **constant time** by maintaining an **additional field** $v.pos$.

$v.pos$ is **-1** when v is **extracted** from H . This also makes test $v \in H$ efficient.



It	$v_1.key/v_1.p$	$v_2.key/v_2.p$	$v_3.key/v_3.p$	$v_4.key/v_4.p$	$v_5.key/v_5.p$	u	$H \text{ root}$
0	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	-	v_2
1	$1/v_2$		$5/v_2$	$5/v_2$	∞/nil	v_2	v_1
2			$5/v_2$	$5/v_2$	∞/nil	v_1	v_4
3			$3/v_4$		$3/v_4$	v_4	v_3
4					$2/v_3$	v_3	v_5
5						v_5	empty

Prim's MST Algorithm: Worst-case Run Time

Assuming using binary min-heap and adjacency lists.

$$n = |V| \quad m = |E|$$

1. *for* loop on Lines #3 to #5: $\mathcal{O}(n)$

2. Building Min-Heap: $\mathcal{O}(n)$

3. Heap Operations: $\mathcal{O}(n \lg n)$

4. Total run time for all ExtractMin calls: $\mathcal{O}(n \lg n)$

5. Total run time for all DecreaseKey calls: $\mathcal{O}(m \lg n)$

Total running time: $\mathcal{O}((n+m) \lg n)$

Since the input graph G is **connected** the run time is:

$$m \geq n-1 \Rightarrow n \in \mathcal{O}(m)$$

$$\mathcal{O}(m \lg n)$$

Overview: Prim's and Kruskal's Algorithms

	Input Graph G	Keep track of Connected Components	Find Minimum Weight Edge
Kruskal's	If G is not connected generates a minimum spanning forest	Disjoint-Sets ADT	Sort all edges according to weight
Prim's	G has to be connected	Keep One Tree plus Isolated Vertices	Use Priority Queue ADT

↓

generates a
minimum spanning Forest
for disconnected graphs
under certain conditions

- After-lecture Readings: Section on "Minimum spanning trees" in the course notes, CLRS Chapter 23
- Problems 23.1-1, 23.2-2 in CLRS.