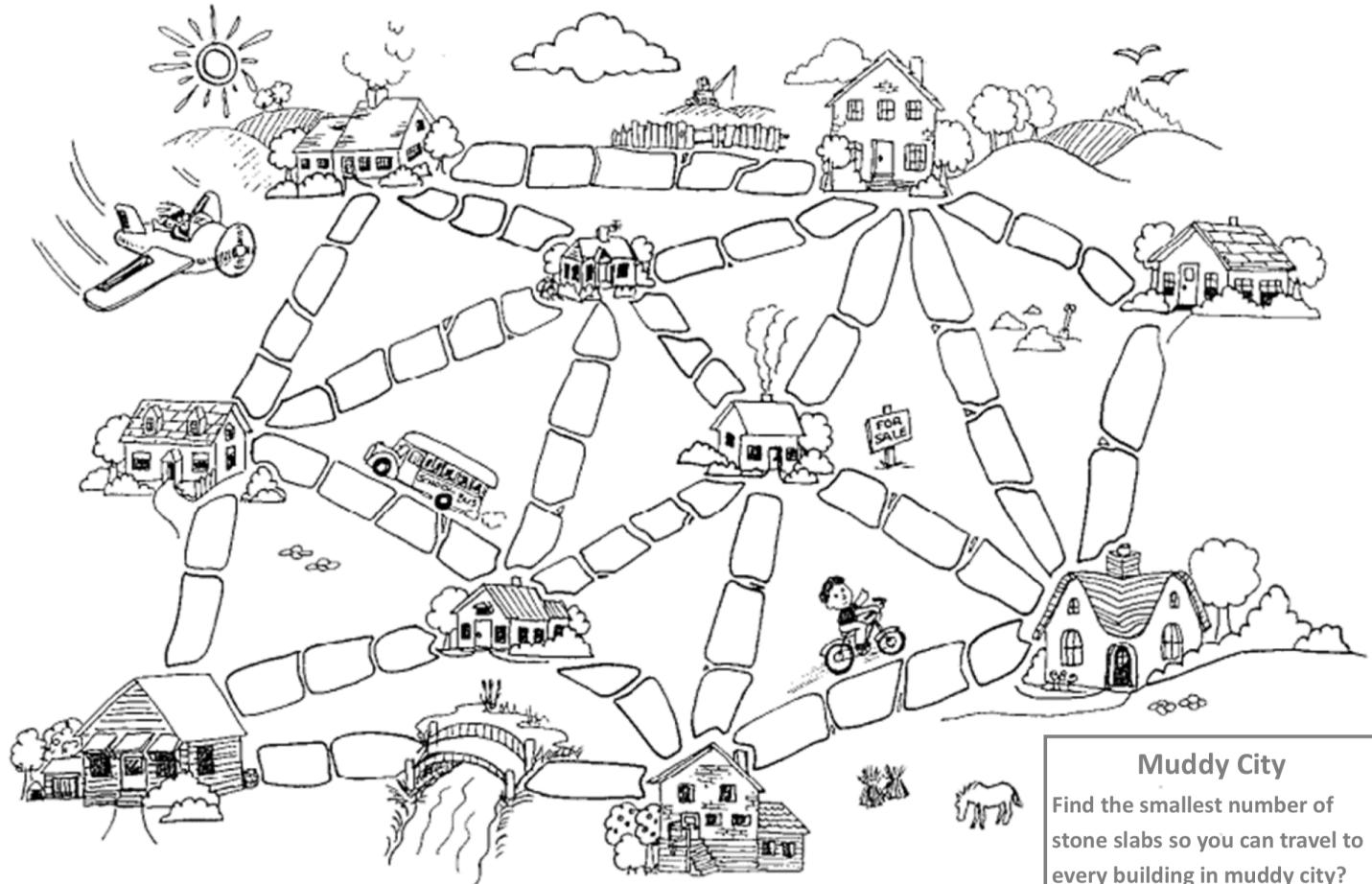


# CSCI 3104: Algorithms

## Lecture 15: Minimum Spanning Trees – continued

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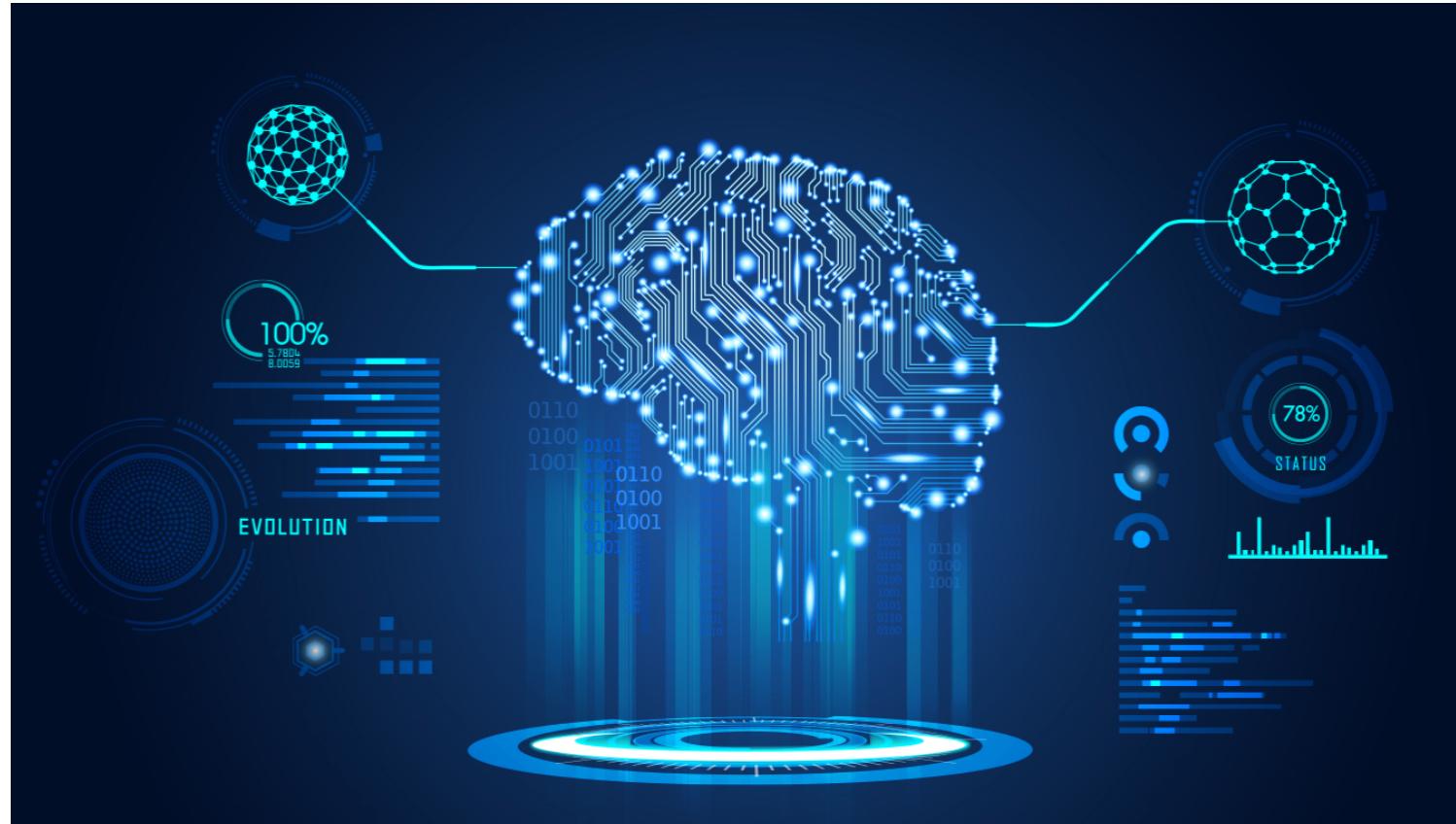


[Image Source](#)

# What will we learn today?

- Minimum Spanning Trees
- Generic Algorithm
- Maximum Flow, Minimum Cut

Intro to Algorithms, CLRS:  
Sections 23.2, 26.1



## Acknowledgements & Citations

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- Material, examples, and definitions taken from CSCI 3104 Week 9 Notes by Professor Aaron Clauset.

## Minimum Spanning Trees – Recap

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A **spanning tree**  $T$  is a set of edges  $E' \subseteq E$ , of size  $|E'| = |V| - 1$ , such that for every pair of vertices  $u, v \in V$ , there exists a path from  $u$  to  $v$  in  $T$ .

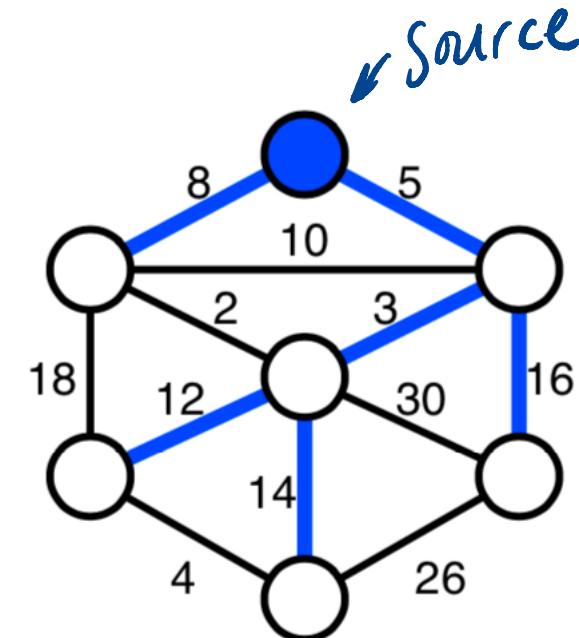
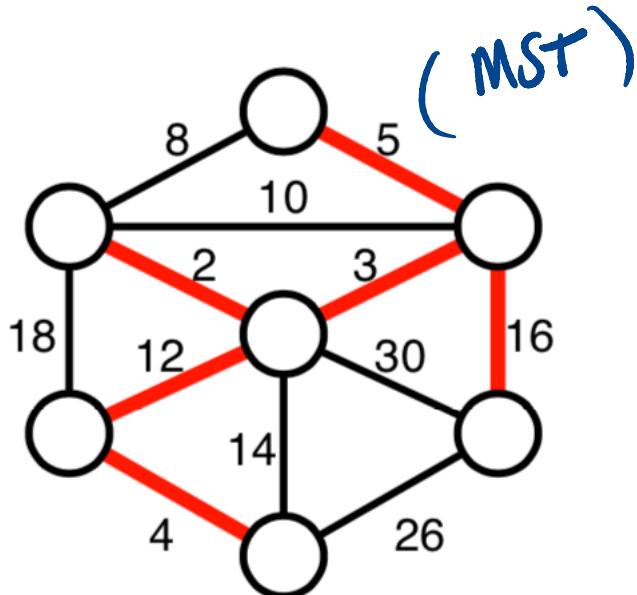
A **minimum spanning tree** (MST) is defined as a spanning tree  $T$  that minimizes the “cost” function.

$$w(T) = \sum_{e \in T} w(e)$$

A minimum spanning tree provides a minimum cost to finding a path between any pair of vertices

# Minimum Spanning Trees – Recap

Example: The red edges below show the Minimum Spanning Tree, while the blue edges show a Single-Source Shortest Path



Single-Source  
Shortest  
Path  
tree  
(SSSP tree)

SSSP tree  $\neq$  MST

# Minimum Spanning Trees – Generic Algorithm

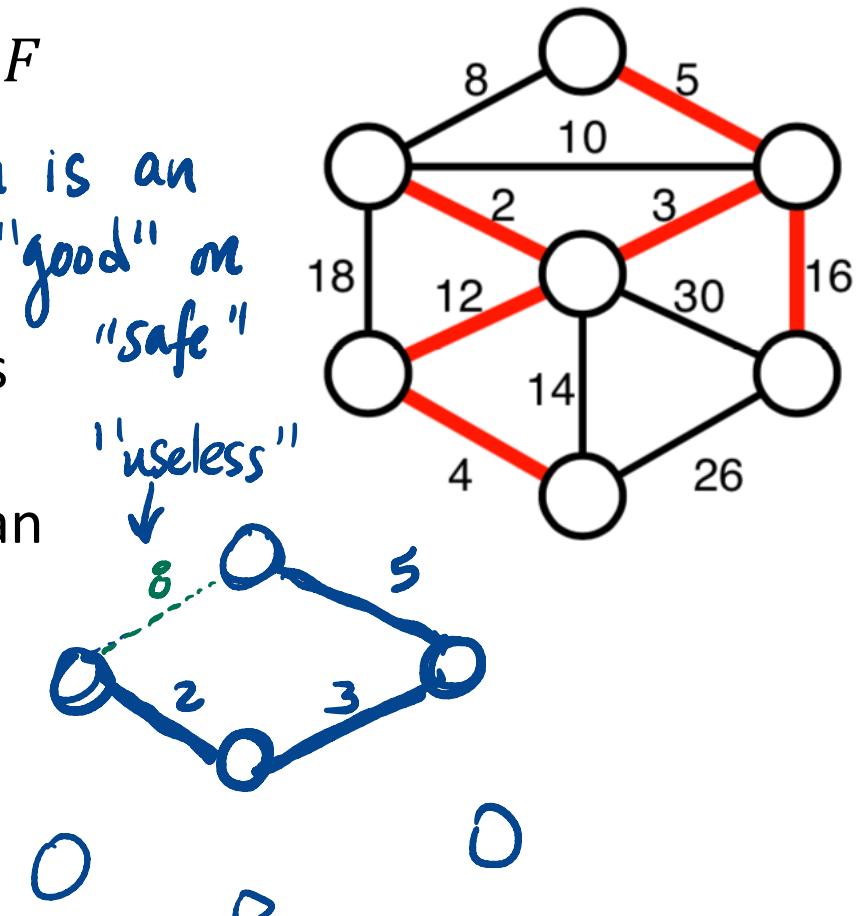
At each step of the algorithm:

- Maintain and update an acyclic subgraph  $F$  on the input graph  $G$ .
- At each step of the algorithm, we add some edges to  $F$

↳ when is an  
edge "good" or  
"safe"

An edge is **useless** if it is not an edge in  $F$  but both of its endpoints are in the same component of  $F$ .

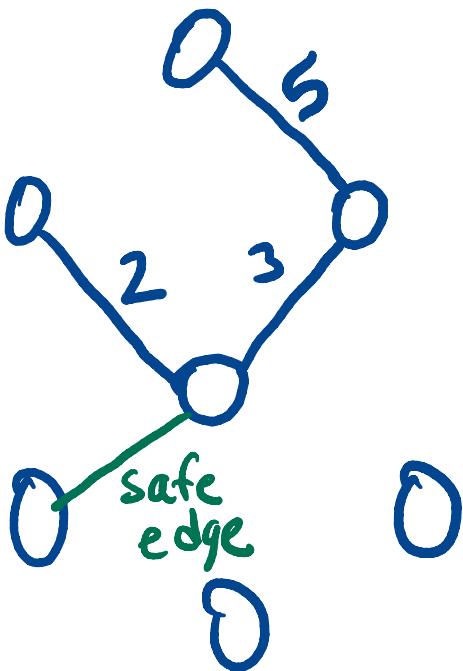
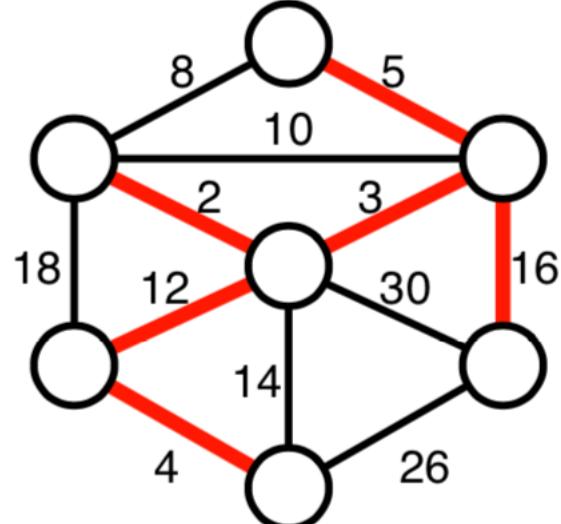
Adding a useless edge to  $F$  would induce a cycle, and can thus never be part of the MST we are building.



# Minimum Spanning Trees – Generic Algorithm

An edge is called **safe** if it is the minimum weight edge among all edges with exactly one endpoint in that component.

- If we assume that edge weights are distinct, a given component can only have one safe edge.



- Edges that are neither safe nor useless are called undecided edges.

# Minimum Spanning Trees – Generic Algorithm

```
Generic-MST(G,w) {  
    F = {}  
    while (F not a spanning tree) {  
        • find edge (u,v) that is safe for F  
        • add (u,v) to F  
    }  
    return F  
}
```

- in each step we add a safe edge
  - Stop when we have no more safe edges
- // initially F contains no edges  
// are we done yet?  
// make a safe choice  
// and grow F  
  
// we are done

→ Prim's

- grow the MST from some initial vertex until we span the graph
- use a min-heap with the edge weights as keys
- check edge to see if its useless or safe

→ Kruskel's

- repeatedly adding the smallest weight edge
- no cycles

# Theory Behind MST Algorithms

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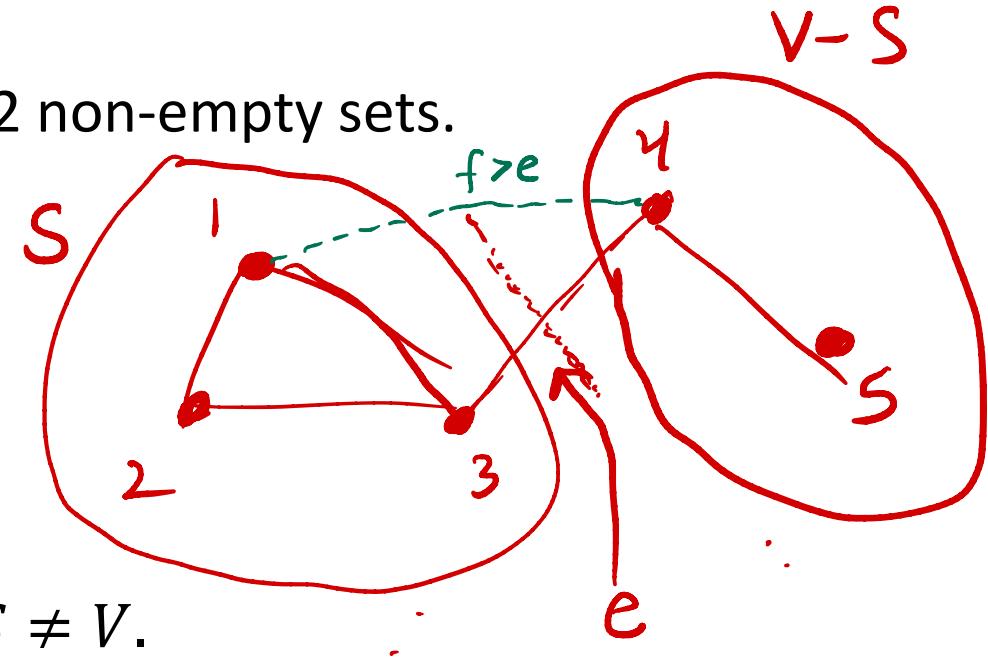
- When is it safe to add an edge to an MST?
- How do we know that the edge selection produces a valid MST?

# Theory Behind MST Algorithms

A **cut** of a graph  $G = (V, E)$  is a partition of  $V$  into 2 non-empty sets.

## Cut Property:

- Assume distinct edge costs.
  - Let  $S$  be a set of vertices that is not empty and  $S \neq V$ .
  - Let edge  $e = (v, w)$  be a minimum cost edge with  $v \in S$  and  $w \in V - S$
- Every MST contains edge  $e$ .



# Theory Behind MST Algorithms

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Example: Prove the Cut Property.

Idea: Exchange Argument

- $T$  - spanning tree

- Identify an edge  $e'$  in  $T$  that is more expensive than  $e$  and exchange  $e$  and  $e'$

# Theory Behind MST Algorithms

Example: Prove the Cut Property.

Proof: Let  $e = (v, w)$  be the minimum cost edge with  $v \in S$  and  $w \in V-S$

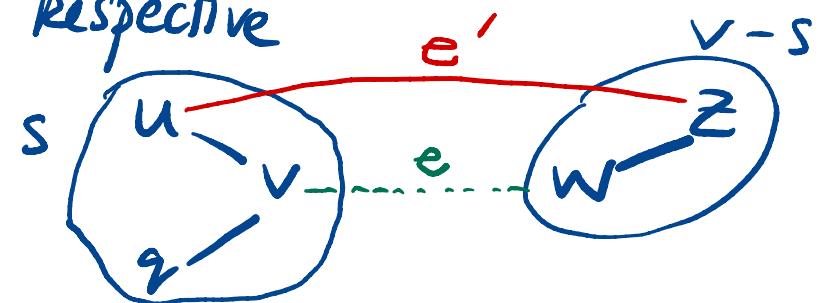
Let  $T$  be a spanning tree that doesn't contain  $e$ .

Note, as a spanning tree  $(V, T)$  is connected and acyclic.

We'll show that  $T$  is not the minimum possible cost.

The ends of  $e$  are in  $S$  and  $V-S$  respectively.

We select an  $e'$  that also has its respective edges in  $S$  and  $V-S$



# Theory Behind MST Algorithms

Example: Prove the Cut Property.

Let  $w' \in S$  and  $v' \in V-S$

such that  $e' = (v', w')$

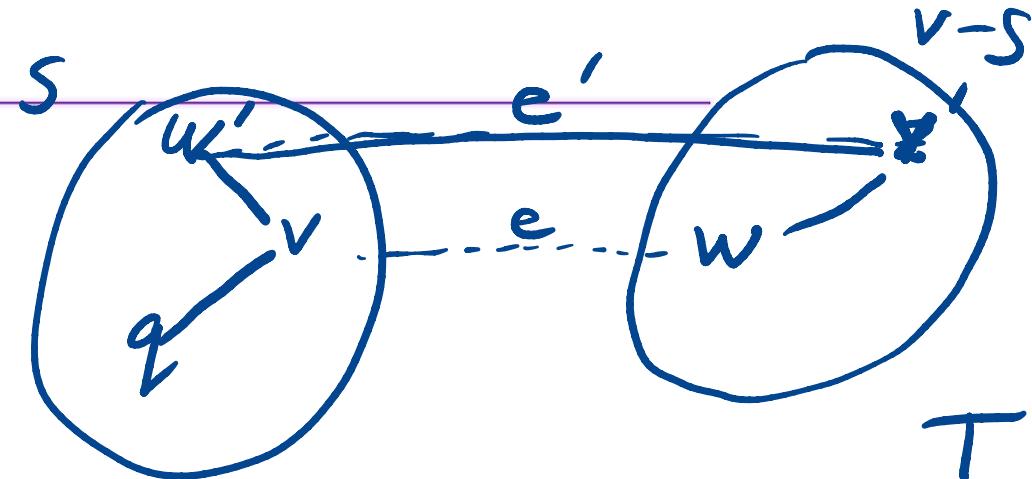
is in the spanning tree  $T$

Exchange edge  $e$  for edge  $e'$  to get edge set  $T'$

where  $T' = T - \{e'\} \cup \{e\}$

Thus  $T'$  contains edge  $e$ .

We know  $\underline{(V, T')}$  is connected because  $(V, T)$  was connected.



# Theory Behind MST Algorithms

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Example: Prove the Cut Property.

Path edge  $e' = (v', w')$  in  $T$  is replaced by  $e = (v, w)$  in  $T'$  where  $v \in S$  and  $w \in V-S$

Since the only edges connecting  $S$  and  $V-S$  were  $e$  and  $e'$  and  $e'$  was removed, we know that  $T'$  is acyclic.

Both  $e$  and  $e'$  have an end in  $S$  and the other end in  $V-S$ .  $e$  is cheaper  $\Rightarrow C_e < C_{e'}$   
 $\Rightarrow$  total cost of  $T' <$  total cost of  $T$

# Theory Behind MST Algorithms

## Cycle Property:

- Let  $c$  be any cycle in  $G$  and let edge  $e = (v, w)$  be the most expensive edge in the cycle  $c$ . Then,  $e$  does not belong to any MST of  $G$ .

~~no proof~~

Pf: Suppose e belongs to some MST

If we delete edge  $e$  from  $T^*$ , then

$T^*$  is now disconnected

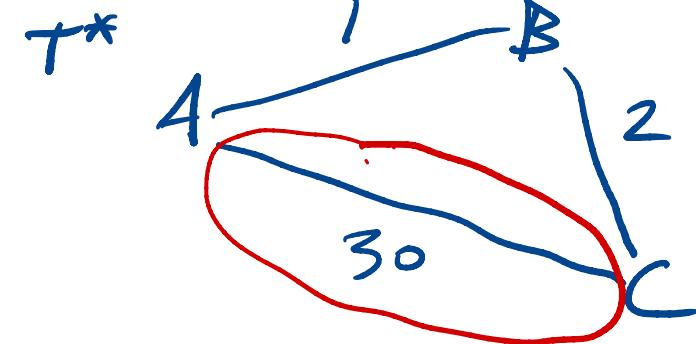
Suppose there is some other edge,

$f$

$T = T^* \cup \{f\} - \{e\}$  is also

a spanning tree. Since  $c_f < c_e$

$\text{cost}(T) < \text{cost}(T^*)$ .  $\Rightarrow$  that  $T^*$  is MST.



$\overline{AC}$  is not part of  
any MST

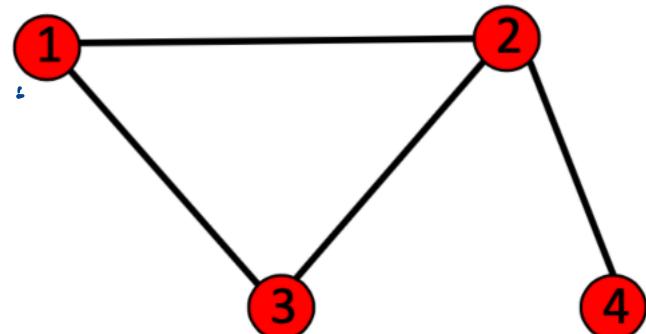
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# Review? – Adjacency Matrices

- An **adjacency matrix** is one way to represent a graph.

Example: Write the adjacency matrix for the following unweighted, undirected graph.



let  $a_{ij} = \begin{cases} 1 & \text{if } e_{ij} \in E \\ 0 & \text{otherwise} \end{cases}$

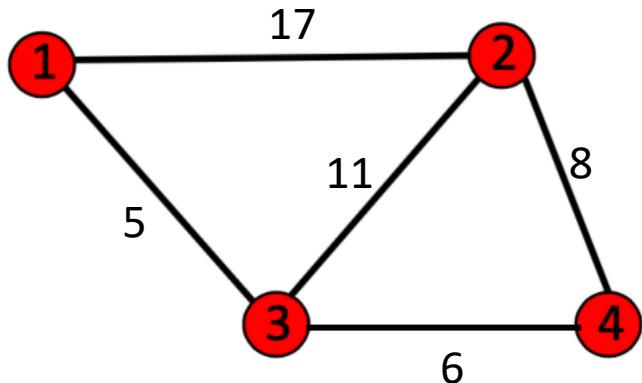
$$A = \begin{bmatrix} & \text{column}_1 & \text{column}_2 & \text{column}_3 & \text{column}_4 \\ \text{Row}_1 & 0 & 1 & 1 & 0 \\ \text{Row}_2 & 1 & 0 & 1 & 1 \\ \text{Row}_3 & 1 & 1 & 0 & 0 \\ \text{Row}_4 & 0 & 1 & 0 & 0 \end{bmatrix}$$

# Review? – Adjacency Matrices

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- An **adjacency matrix** is one way to represent a graph.

Example: Write the adjacency matrix for the following weighted, undirected graph.



$$a_{ij} = \begin{cases} w_{ij} & \text{if } e_{ij} \in E \\ 0 & \text{otherwise} \end{cases}$$

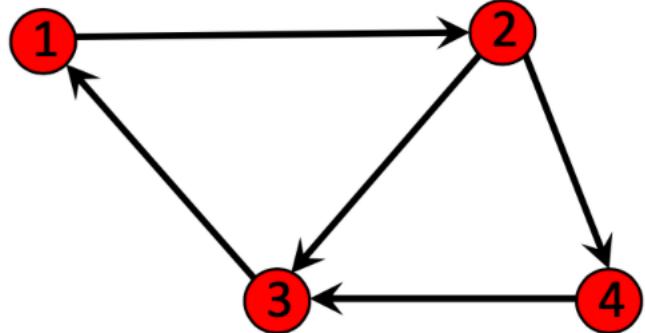
$$A = \begin{bmatrix} 0 & 17 & 5 & 0 \\ 17 & 0 & 11 & 8 \\ 5 & 11 & 0 & 6 \\ 0 & 8 & 6 & 0 \end{bmatrix}$$

# Review? – Adjacency Matrices

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- An *adjacency matrix* is one way to represent a graph.

Example: Write the adjacency matrix for the following ~~undirected~~ graph.



$$A = \begin{bmatrix} & 1 & 2 & 3 & 4 \\ 1 & 0 & 1 & 0 & 0 \\ 2 & 0 & 0 & 1 & 1 \\ 3 & 1 & 0 & 0 & 0 \\ 4 & 0 & 0 & 1 & 0 \end{bmatrix}$$

↑  
not  
specified

# Maximum Flow

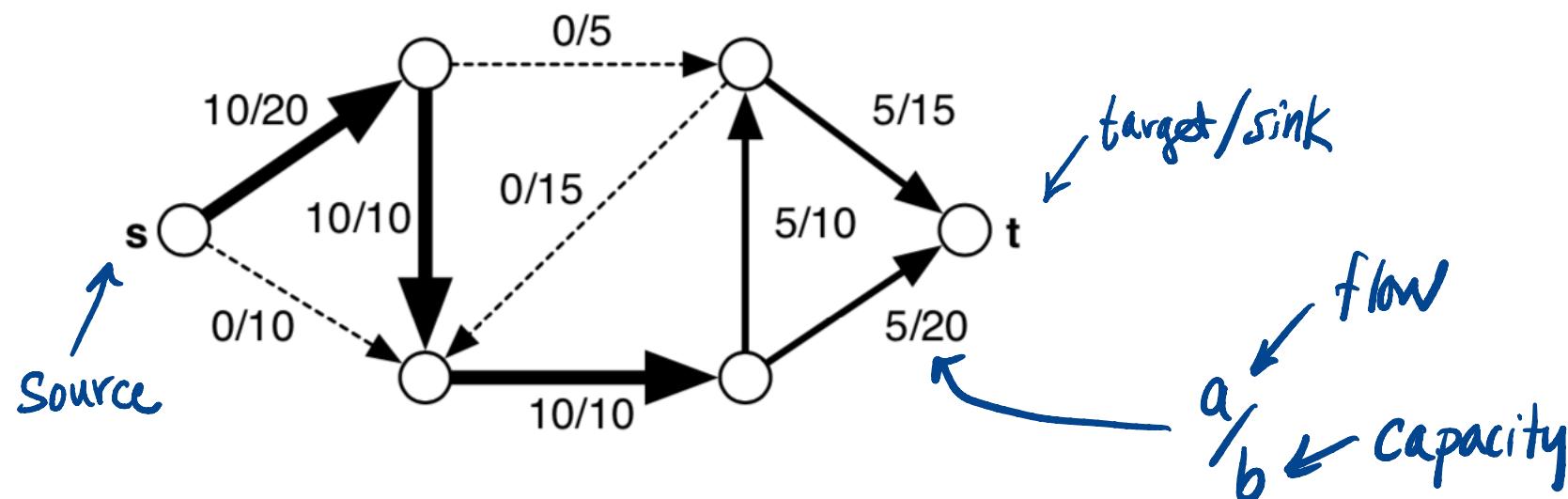
In max-flow:

❖ the input is:

- a directed graph  $G = (V, E)$ ,
- an edge weight function that represents the capacities of the edges,
- and a pair of vertices that represent the source and target

*Sink*

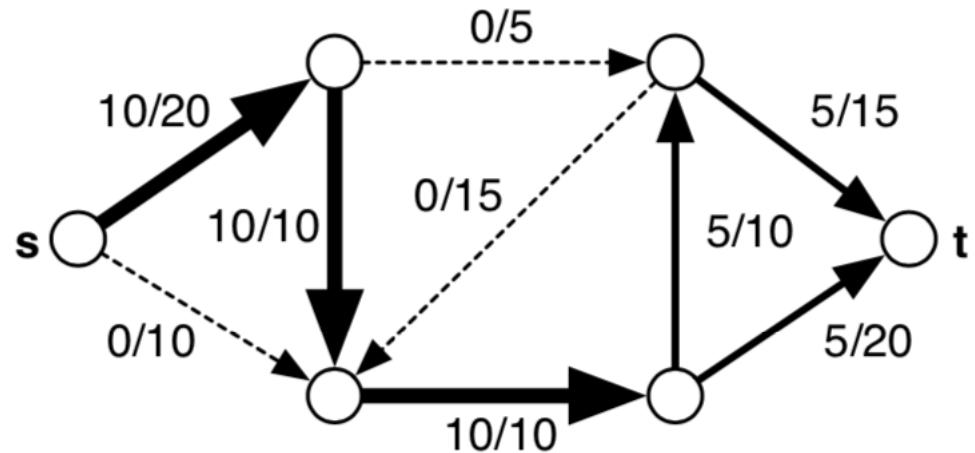
❖ The output is a flow function.



# Maximum Flow

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**The maximum flow problem:** Given a graph  $G$ , capacity function  $c$ , and choices  $s, t$ , compute a feasible  $(s, t)$  – flow on  $G$  whose value  $|f|$  is maximized.



- Flow can only be created at the source  $s$
- Flow can only be consumed at the target  $t$

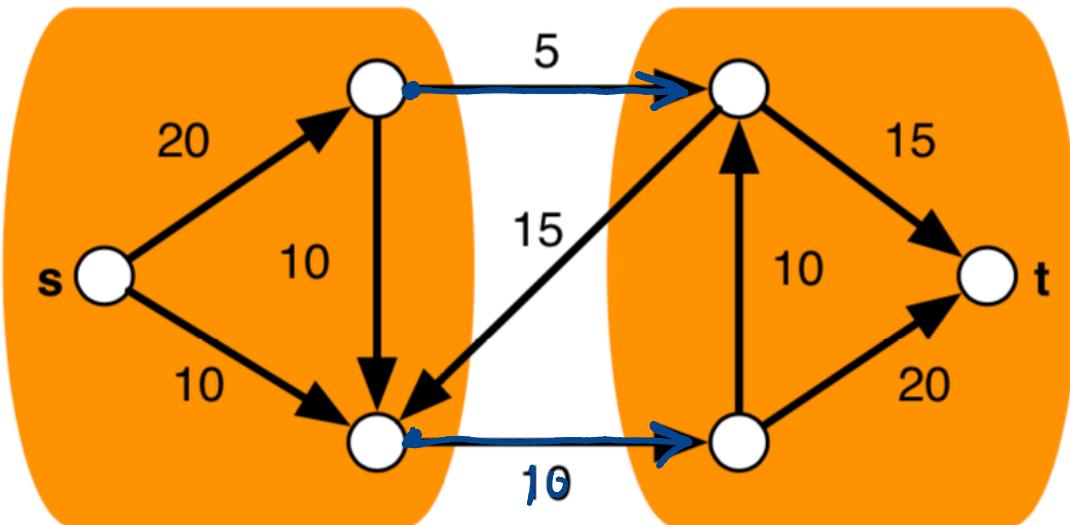
A flow is **feasible** with respect to the edge capacities  $c$  if for every edge  $e \in E$ , the capacity of  $e$  is not exceeded by the flow on  $e$ ;  $f(e) \leq c(e)$



# Minimum Cut

**The minimum cut problem:** Given a graph  $G$ , a capacity function  $c$ , and choices  $s, t$ , find an  $(s, t) - \text{cut}$  whose capacity  $\|S, T\|$  is minimized.

A  $(s, t) - \text{cut}$  on  $G$  is a bipartitioning of the vertices  $V$  into subsets  $S$  and  $T$ , with  $s \in S$  and  $t \in T$



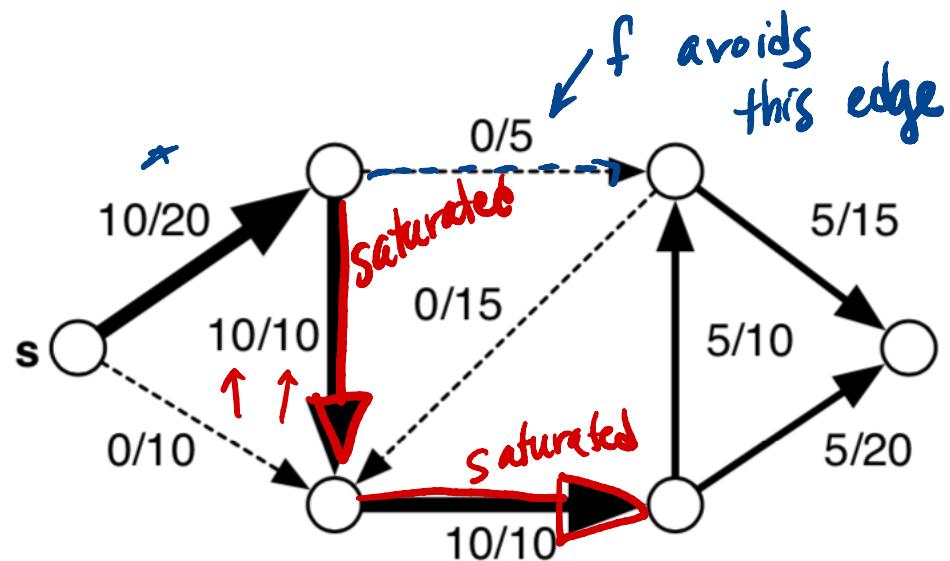
The **capacity** of the cut is the total capacity of the edges “in” the cut, which is defined as edges that start in  $S$  and end in  $T$ .

$$\|S, T\| = 15$$

# Max-Flow Min-Cut

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- If  $f(e) = c(e)$ , then we say that  $f$  **saturates** the edge  $e$ .
- If  $f(e) = 0$ , then we say that  $f$  **avoids** the edge  $e$



# Max-Flow Min-Cut

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***Max-Flow Min-Cut Theorem:*** The value of the max flow equals the capacity of the min cut.

A simple greedy approach:

- Saturate the largest capacity path from  $s$  to  $t$ , recurse until no more flow can be pushed through.