

Week 11. Lecture Notes

Topics: Disjoint set data structure
Union-Find
Augmented disjoint set data structure
Network flow

Disjoint-set data structure (Union-Find)

Problem:

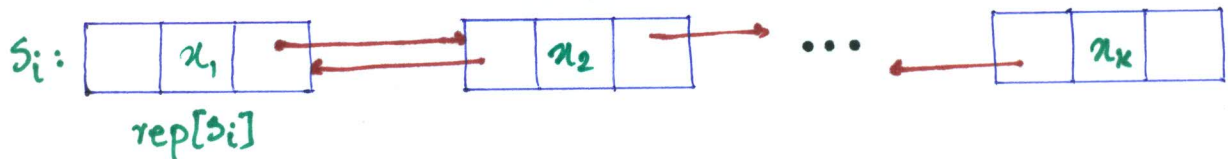
Maintain a dynamic collection of pairwise-disjoint sets $S = \{S_1, S_2, \dots, S_r\}$. Each set S_i has one element distinguished as the representative element, $\text{rep}[S_i]$

Must support 3 operations:

- MAKE-SET(x): adds new set $\{x\}$ to S with $\text{rep}[\{x\}] = x$ (for any $x \notin S_i$ for all i)
- UNION(x, y): replaces sets S_x, S_y with $S_x \cup S_y$ in S for any x, y in distinct sets S_x, S_y ,
- FIND-SET(x): returns representative $\text{rep}[S_x]$ of set S_x containing element x .

Simple linked-list solution

Store each set $S_i = \{x_1, x_2, \dots, x_k\}$ as an (unordered) doubly linked list. Define representative element $\text{rep}[S_i]$ to be the front of the list, x_1 .

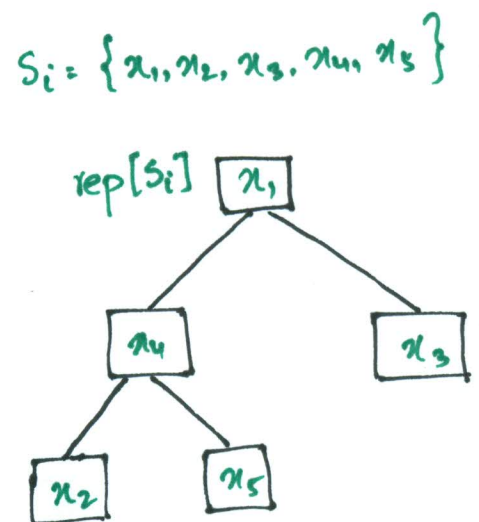


- $\text{MAKE-SET}(x)$ initializes x as a lone node — $\Theta(1)$
- $\text{FIND-SET}(x)$ walks left in the list containing x until it reaches the front of the list — $\Theta(n)$
- $\text{UNION}(x, y)$ concatenates the lists containing x and y , leaving rep. as $\text{FIND-SET}[x]$ — $\Theta(n)$

Simple balanced-tree solution

Store each set $S_i = \{x_1, x_2, \dots, x_k\}$ as a balanced tree (ignoring keys). Define representative element $\text{rep}[S_i]$ to be the root of the tree.

- $\text{MAKE-SET}(x)$ initializes x as a lone node — $\Theta(1)$
- $\text{FIND-SET}(x)$ walks up the tree containing x until it reaches the root — $\Theta(\log n)$
- $\text{UNION}(x, y)$ concatenates the trees containing x and y changing rep. — $\Theta(\log n)$



Plan of attack

We will build a simple disjoint union data structure that, in an amortized sense, performs significantly better than $\Theta(\lg n)$ per operation, even better than $\Theta(\lg \lg n)$, $\Theta(\lg \lg \lg n)$, etc, but not quite $\Theta(1)$.

To reach this goal, we will introduce two key tricks. Each trick converts a trivial $\Theta(n)$ solution into a simple $\Theta(\lg n)$ amortized solution. Together, the two tricks yield a much better solution.

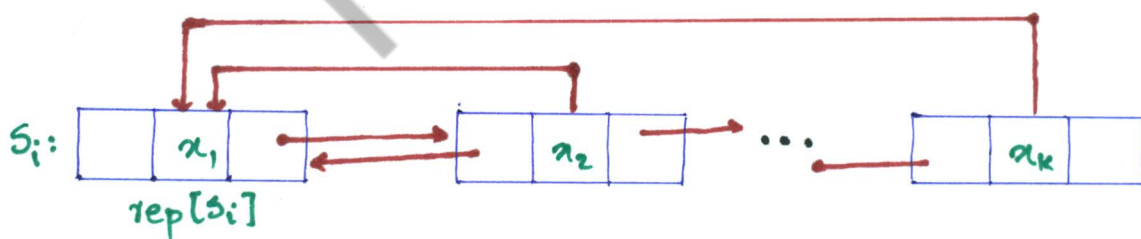
First trick arises in an augmented linked list.

Second trick arises in a tree structure.

Augmented linked-list solution

Store set $S_i = \{x_1, x_2, \dots, x_k\}$ as unordered doubly linked list. Define $\text{rep}[S_i]$ to be front of the list, x_1 .

Each element x_j also stores pointer $\text{rep}[x_j]$ to $\text{rep}[S_i]$



- $\text{FIND-SET}(x)$ returns $\text{rep}[x]$ — $\Theta(1)$
- $\text{UNION}(x, y)$ concatenates the lists containing x and y and updates the rep pointers for all elements in the list containing y . — $\Theta(n)$

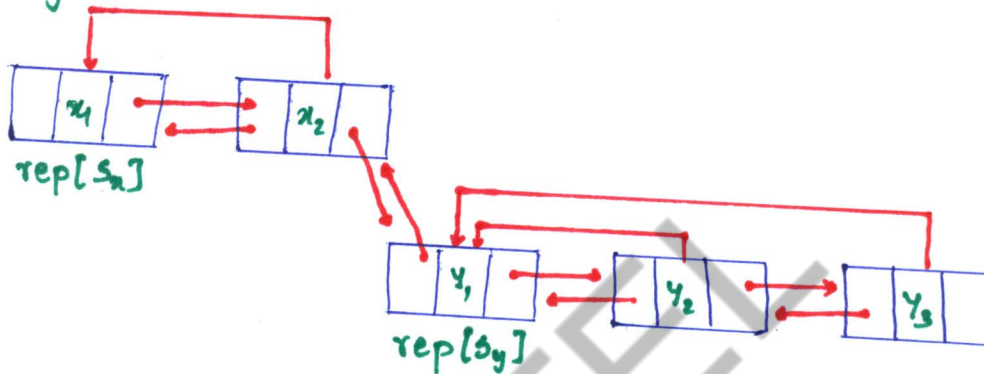
Example of augmented linked-list solution

Each element x_i stores pointer $\text{rep}[x_i]$ to $\text{rep}[s_i]$.

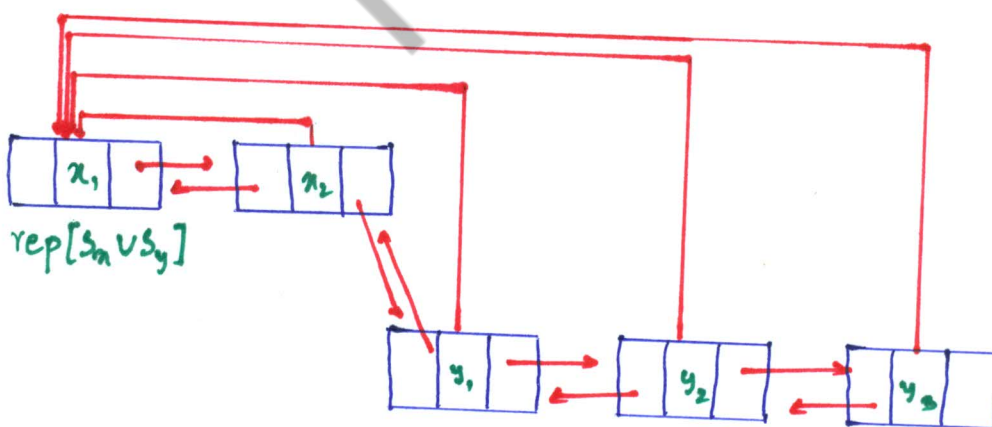
UNION(x, y)

- Concatenates the lists containing x and y , and
- updates the rep pointers for all elements in the list containing y .

$S_x \cup S_y$:



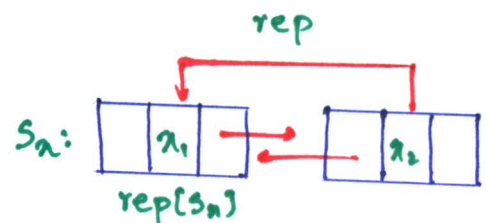
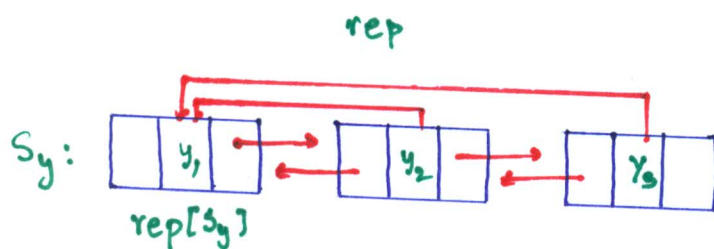
$S_x \cup S_y$:



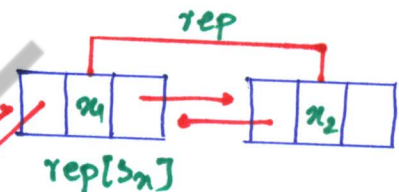
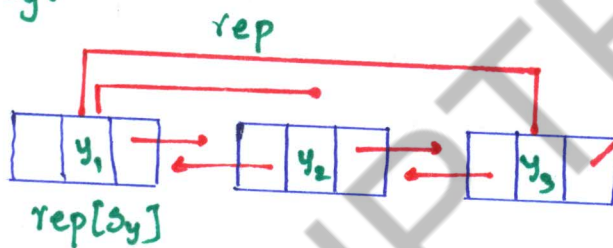
Alternative Concatenation

UNION(x, y) could instead

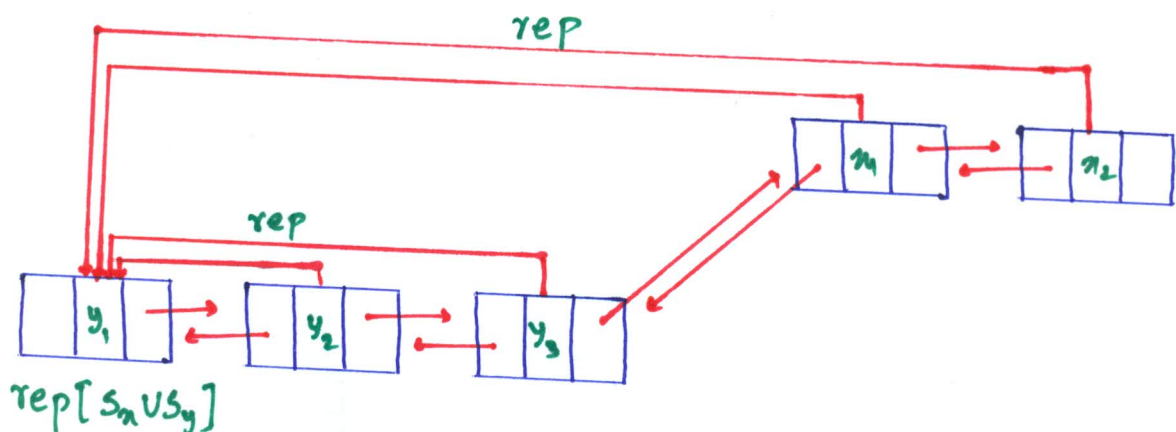
- concatenate the lists containing y and x , and
- update the rep pointers for all elements in the list containing x .



$S_x \cup S_y$:



$S_x \cup S_y$:



Trick 1: Smaller into larger

To save work, concatenate smaller list into the end of the larger list. Cost = $\Theta(\text{length of smaller list})$

Augment list to store its weight (#elements)

Let " n " denote the overall number of elements (equivalently, the number of MAKE-SET operations).

Let " m " denote the total number of operations

Let " f " denote the number of FIND-SET operations.

THEOREM: Cost of all UNION's is $O(n \lg n)$.

Corollary: Total cost is $O(m + n \lg n)$.

Analysis of Trick 1

To save work, concatenate smaller list into the end of the larger list. Cost = $\Theta(1 + \text{length of smaller list})$

Theorem: Total cost of UNION's is $O(n \lg n)$

Proof:

Monitor an element x and set S_x containing it. After initial MAKE-SET(x), weight $[S_x] = 1$. Each time S_x is united with S_y , weight $[S_y] \geq \text{weight}[S_x]$, pay 1 to update $\text{rep}[x]$, and weight $[S_x]$ at least doubles (increasing by weight $[S_y]$).

Each time S_y is united with smaller set S_x , pay nothing, and weight $[S_x]$ only increases.

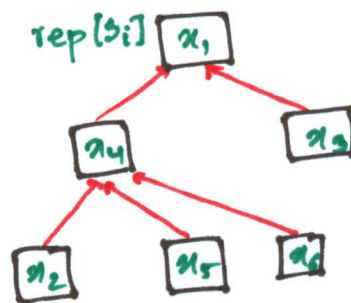
Thus pay $\leq \lg n$ for x

Representing sets as trees

Store each set $S_i = \{x_1, x_2, \dots, x_k\}$ as an unordered, potentially unbalanced, not necessarily binary tree, storing only parent pointers. $\text{rep}[S_i]$ is the tree root.

$$S_i = \{x_1, x_2, x_3, x_4, x_5, x_6\}$$

- $\text{MAKE-SET}(x)$ initializes x as a lone node $\rightarrow \Theta(1)$
- $\text{FIND-SET}(x)$ walks up the tree containing x until it reaches the root $\rightarrow \Theta(\text{depth}[x])$
- $\text{UNION}(x, y)$ concatenates the trees containing x and y



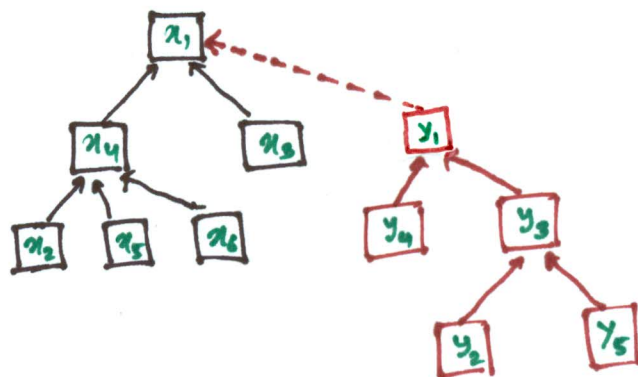
Trick 1 adapted to trees

$\text{UNION}(x, y)$ can use a simple concatenation strategy:
Make root $\text{FIND-SET}(y)$ a child of root $\text{FIND-SET}(x)$.

$$\Rightarrow \text{FIND-SET}(y) = \text{FIND-SET}(x)$$

We can adapt Trick 1 to this context also:

Merge tree with smaller weight into tree with larger weight.



Height of tree increases only when its size doubles, so height is logarithmic in weight.

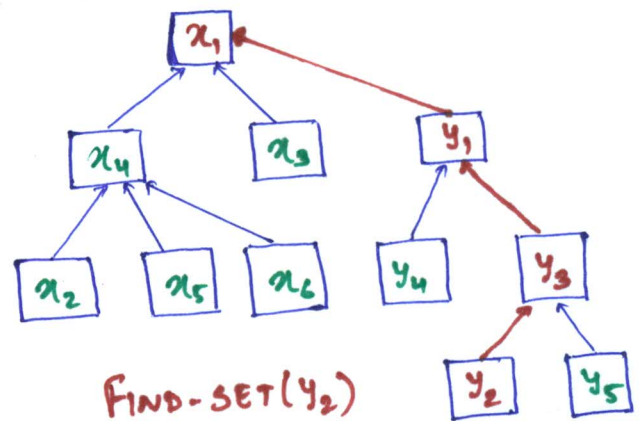
Thus total cost is $O(m + f \lg n)$

Trick 2: Path compression

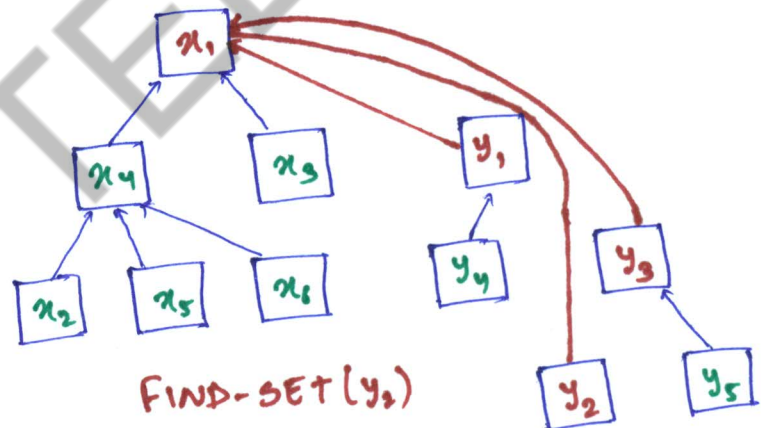
When we execute a FIND-SET operation and walk up a path p to the root, we know the representatives for all nodes on path p .

Path compression makes all of those nodes direct children of the root.

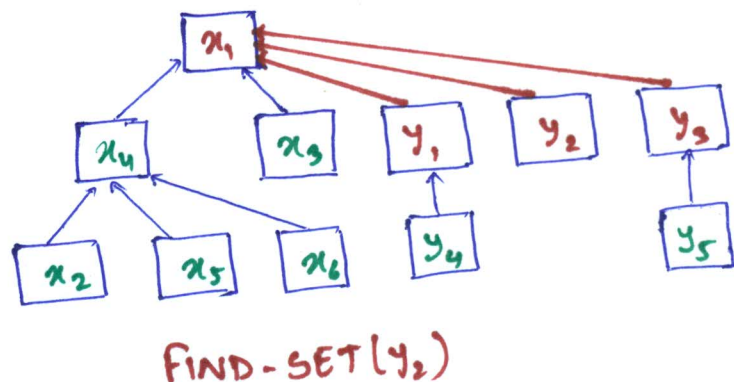
Cost of $\text{FIND-SET}(x)$ is still $\Theta(\text{depth}[x])$



Cost of $\text{FIND-SET}(x)$ is still $\Theta(\text{depth}[x])$



Cost of $\text{FIND-SET}(x)$ is still $\Theta(\text{depth}[x])$



Analysis of Trick 2 alone

Theorem:

Total cost of FIND-SET is $O(m \lg n)$

Proof:

Amortization by potential function.

The weight of a node x is # nodes in its subtree.

Define $\Phi(x_1, \dots, x_n) = \sum_i \lg \text{weight}[x_i]$

UNION (x_i, x_j) increases potential of root FIND-SET(x_i),
by at most $\lg \text{weight}[\text{root FIND-SET}(x_j)] \leq \lg n$.

Each step down $p \rightarrow c$ made by FIND-SET(x_i),
except the first, moves c 's subtree out of p 's subtree.

Thus if $\text{weight}[c] \geq \frac{1}{2} \text{weight}[p]$, Φ decreases by ≥ 1 .

paying for the step down. There can be at most
 $\lg n$ steps $p \rightarrow c$ for which $\text{weight}[c] < \frac{1}{2} \text{weight}[p]$.

Theorem:

If all UNION operations occur before all FIND-SET operations, then total cost is $O(mn)$

Proof:

If a FIND-SET operation traverses a path with k nodes, costing $O(k)$ time, then $k-2$ nodes are made new children of the root.

This change can happen only once for each of the n elements.

So, the total cost of FIND-SET is $O(f + n)$.

Ackermann's function A

$$\text{Define } A_k(j) = \begin{cases} j+1, & \text{if } k=0 \\ A_{k-1}^{(j+1)}(j), & \text{if } k \geq 1 \end{cases} \quad \text{— iterate } j+1 \text{ times}$$

$$A_0(j) = j+1$$

$$A_0(1) = 2$$

$$A_1(j) \sim 2j$$

$$A_1(1) = 3$$

$$A_2(j) \sim 2j \cdot 2^j > 2^j$$

$$A_2(1) = 7$$

$$A_3(j) > 2^{2^2 \dots 2^j} \quad \left. \begin{matrix} \vdots \\ 2^j \end{matrix} \right\} j$$

$$A_3(1) = 2047$$

$$A_4(1) \text{ is a lot bigger. } A_4(1) > 2^{2^2 \dots 2^{2047}} \quad \left. \begin{matrix} \vdots \\ 2^{2047} \end{matrix} \right\} 2048$$

Define $\alpha(n) = \min \{k: A_k(1) \geq n\} \leq 4$ for practical n

Analysis of Tricks 1 + 2

Theorem:

In general, total cost is $O(m \alpha(n))$.

Application: Dynamic Connectivity

Suppose a graph is given to us incrementally by

ADD-VERTEX(v)

ADD-EDGE(u, v)

and we want to support connectivity queries:

CONNECTED(u, v):

Are u and v in the same connected component?

For example, we want to maintain a spanning forest, so we check whether each new edge connects a previously disconnected pair of vertices.

Sets of vertices represent connected components.

Suppose a graph is given to us incrementally by

ADD-VERTEX(v) - MAKE-SET(v)

ADD-EDGE(u, v) - if not CONNECTED(u, v)
then UNION(u, v)

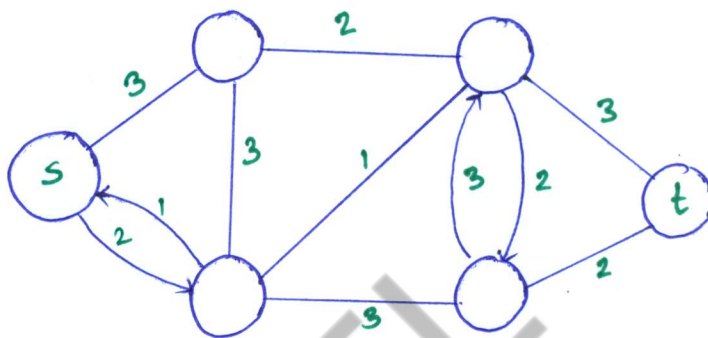
and

CONNECTED(u, v): FIND-SET(u) = FIND-SET(v)

Flow networks

Definition: A flow network is a directed graph $G=(V,E)$ with two distinguished vertices: a source s and a sink t . Each edge $(u,v) \in E$ has a non-negative capacity $c(u,v)$. If $(u,v) \notin E$, then $c(u,v) = 0$.

Example:



Definition: A positive flow on G is a function $p: V \times V \rightarrow \mathbb{R}$ satisfying the following:

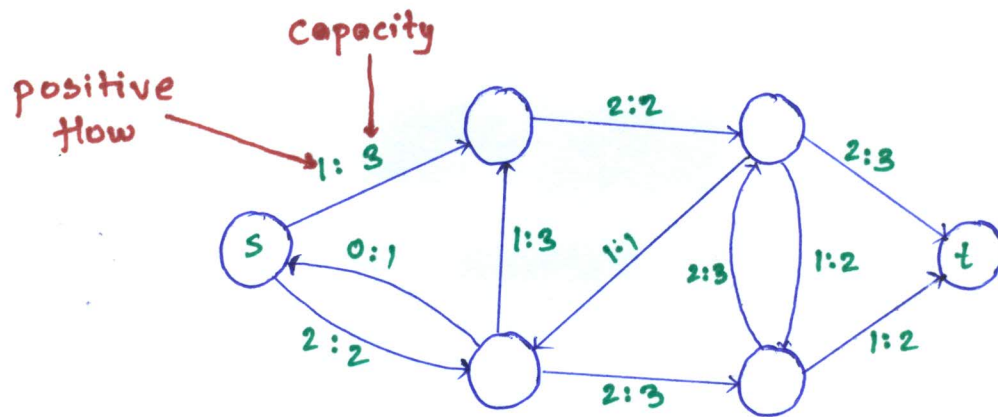
- Capacity constraint: for all $u,v \in V$
 $0 \leq p(u,v) \leq c(u,v)$
- Flow conservation: for all $u \in V - \{s,t\}$,

$$\sum_{v \in V} p(u,v) - \sum_{v \in V} p(v,u) = 0$$

The value of a flow is the net flow out of the source:

$$\sum_{v \in V} p(s,v) - \sum_{v \in V} p(v,s)$$

A flow on a network



Flow conservation (like Kirchoff's current law):

- Flow into u is $2+1=3$
- Flow out of u is $0+1+2=3$

The value of this flow is $1-0+2=3$.

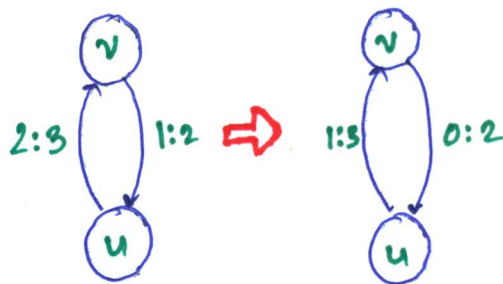
Maximum-flow problem: Given a flow network G , find a flow of maximum value in G .

In the above figure, value of maximum flow = 4.

Flow Cancellation

Without loss of generality, positive flow goes either from u to v , or from v to u , but not both.

The capacity constraint and flow conservation are preserved by this transformation



Net flow from u to v in both cases is 1

INTUITION:

View flow as a rate, not a quantity.

A notational simplification

IDEA: Work with the net flow between two vertices, rather than with the positive flow.

Definition: A (net) flow on G is a function $f: V \times V \rightarrow \mathbb{R}$ satisfying the following:

- Capacity constraint: For all $u, v \in V$,

$$f(u, v) \leq c(u, v)$$

- Flow conservation: For all $u \in V - \{s, t\}$

$$\sum_{v \in V} f(u, v) = 0 \quad \left. \begin{array}{l} \text{one assumption} \\ \text{instead of two} \end{array} \right\}$$

- Skew-symmetry: For all $u, v \in V$

$$f(u, v) = -f(v, u)$$

Equivalence of Definitions

Theorem: The two definitions are equivalent

Proof:

$$\text{Let } f(u, v) = p(u, v) - p(v, u)$$

Capacity constraint: Since $p(u, v) \leq c(u, v)$ and $p(v, u) \geq 0$, we have $f(u, v) \leq c(u, v)$.

Flow conservation:

$$\begin{aligned} \sum_{v \in V} f(u, v) &= \sum_{v \in V} (p(u, v) - p(v, u)) \\ &= \sum_{v \in V} p(u, v) - \sum_{v \in V} p(v, u) \end{aligned}$$

Skew symmetry: $f(u,v) = p(u,v) - p(v,u)$

$$= -(p(v,u) - p(u,v))$$

$$= -f(v,u)$$

Next, consider

$$p(u,v) = \begin{cases} f(u,v), & \text{if } f(u,v) > 0 \\ 0, & \text{if } f(u,v) \leq 0 \end{cases}$$

Capacity constraint: By definition $p(u,v) \geq 0$. Since $f(u,v) \leq c(u,v)$, it follows that $p(u,v) \leq c(u,v)$

Flow conservation: If $f(u,v) > 0$, then $p(u,v) - p(v,u) = f(u,v)$.
 If $f(u,v) \leq 0$, then $p(u,v) - p(v,u) = -f(v,u) = f(u,v)$
 (by skew symmetry)

Therefore,

$$\sum_{v \in V} p(u,v) - \sum_{v \in V} p(v,u) = \sum_{v \in V} f(u,v)$$

Notation

Definition: The value of a flow f , denoted by $|f|$ is given by

$$|f| = \sum_{v \in V} f(s,v) = f(s,V)$$

Implicit summation notation:

A set used in an arithmetic formula represents a sum over the elements of the set.

Example:

flow conservation:

$$f(u,v) = 0 \text{ for all } u \in V - \{s,t\}.$$

Simple properties of flow

Lemma:

$$f(x, x) = 0$$

$$f(x, y) = -f(y, x)$$

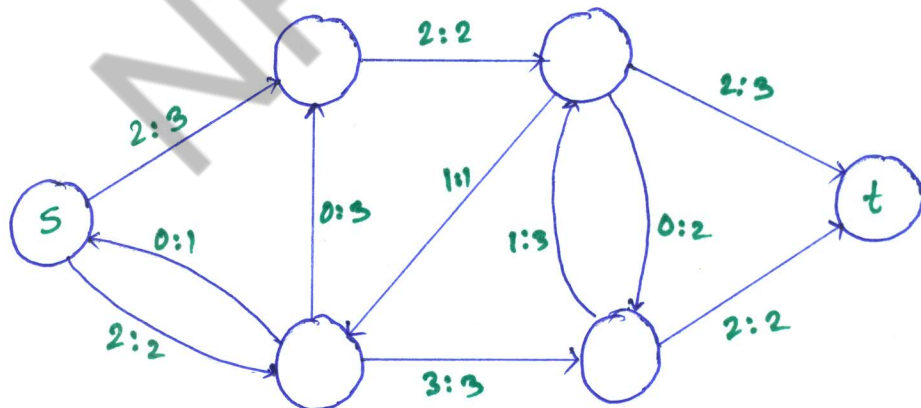
$$f(x \cup y, z) = f(x, z) + f(y, z) \text{ if } x \cap y = \emptyset$$

Theorem

$$|f| = f(v, t)$$

Proof: $|f| = f(s, v)$
 $= f(v, v) - f(v-s, v)$ Omit braces
 $= f(v, v-s)$
 $= f(v, t) + f(v, v-s-t)$
 $= f(v, t)$

Flow into the sink



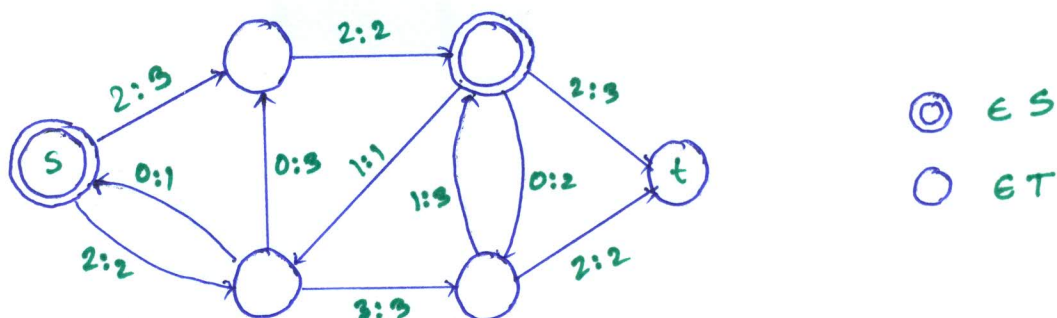
$$|f| = f(s, v) = 4$$

$$f(v, t) = 4.$$

Cuts

A **cut** (S, T) of a flow network $G = (V, E)$ is a partition of V such that $s \in S$ and $t \in T$.

If f is a flow on G , then the flow across the cut is $f(S, T)$



$$f(S, T) = (2+2) + (-2+1-1+2) = 4$$

Another characterization of flow value

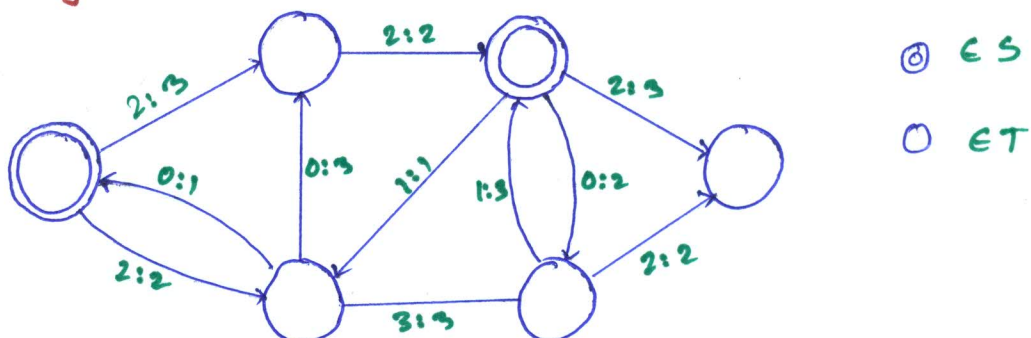
Lemma: For any flow f and any cut (S, T) $|f| = f(S, T)$.

Proof:

$$\begin{aligned} f(S, T) &= f(s, V) - f(s, s) \\ &= f(s, V) \\ &= f(s, V) - f(s - s, V) \\ &= f(s, V) = |f| \end{aligned}$$

Capacity of a cut

capacity of a cut (S, T) is $c(S, T)$



$$\begin{aligned} c(S, T) &= (3+2) + (1+2+3) \\ &= 11 \end{aligned}$$

Upper bound on the maximum flow value

Theorem:

The value of any flow is bounded above by the capacity of any cut.

Proof:

$$\begin{aligned} |f| &= f(S, T) \\ &= \sum_{u \in S} \sum_{v \in T} f(u, v) \\ &\leq \sum_{u \in S} \sum_{v \in T} c(u, v) \\ &= \underline{\underline{c(S, T)}} \end{aligned}$$

Residual Network

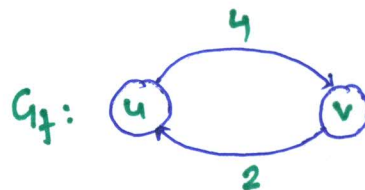
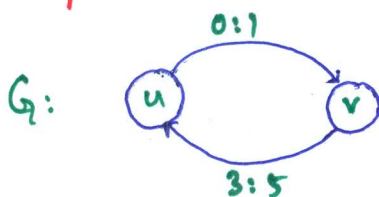
Definition:

Let f be a flow on $G = (V, E)$. The residual network $G_f(V, E_f)$ is the graph with strictly positive residual capacities:

$$c_f(u, v) = c(u, v) - f(u, v) > 0.$$

Edges in E_f admit more flow.

Example:



Lemma:

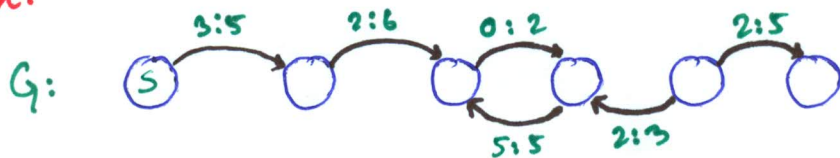
$$|E_f| \leq 2|E|$$

Augmenting Paths

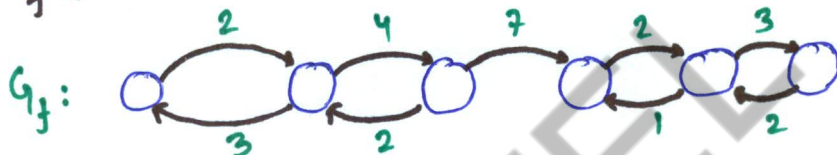
Definition: Any path from s to t in G_f is an augmenting path in G with respect f . The flow value can be increased along an augmenting path p by

$$c_f(p) = \min_{(u,v) \in p} \{c_f(u,v)\}$$

Ex:



$$c_f(p) = 2$$



Max-flow, min cut theorem

Theorem:

The following are equivalent:

1. f is a maximum flow
2. f admits no augmenting paths
3. $|f| = c(s,T)$ for some cut (s,T)