

Algorithms and Data Structures 2020/21 Week 7 Solutions

1. Given a flow network $\mathcal{N} = (\mathcal{G} = (V, E), c, s, t)$, let f_1 and f_2 be two flows in \mathcal{N} (ie, satisfying the three flow properties wrt \mathcal{N}). The *flow sum* $f_1 + f_2$ is the function from $V \times V$ to \mathbb{R} defined by:

$$(f_1 + f_2)(u, v) = f_1(u, v) + f_2(u, v)$$

for all $u, v \in V$.

Which of the three flow properties (wrt \mathcal{N}) will $f_1 + f_2$ satisfy, and which will it violate?

Answer: The three properties are *capacity constraints*, *skew-symmetry*, and *flow conservation*.

Capacity constraints: $f_1 + f_2$ might *violate* the capacity constraints. As an example, consider the network of question 2. Let f_1 be the flow shown in question 2. Let f_2 be the flow that ships 4 units along the path $s \rightarrow x \rightarrow y \rightarrow t$. Then if we add these flows directly as prescribed in this question, we will (for example) define

$$(f_1 + f_2)(y, t) = f_1(y, t) + f_2(y, t) = 4 + 4 = 8.$$

This certainly breaks the capacity constraint for (y, t) which is 4.

Skew-symmetry: $f_1 + f_2$ will *satisfy* skew-symmetry. We know f_1 and f_2 individually satisfy skew-symmetry, because they are flows. Therefore for any (u, v) , we have

$$(f_1 + f_2)(u, v) = f_1(u, v) + f_2(u, v) = -f_1(v, u) - f_2(v, u) = -(f_1 + f_2)(v, u),$$

as required (using the defn of $f_1 + f_2$ and the skew-symmetry property for f_1, f_2).

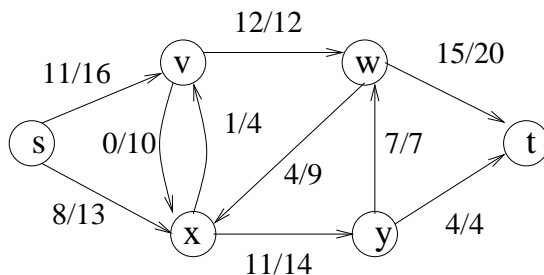
Flow conservation: $f_1 + f_2$ will *satisfy* flow conservation. Flow conservation for a flow f states that for all $u \in V \setminus \{s, t\}$, we have $\sum_{v \in V} f(u, v) = 0$. We know this holds individually for f_1, f_2 . Let $u \in V \setminus \{s, t\}$. Then we can write

$$\sum_{v \in V} (f_1 + f_2)(u, v) = \sum_{v \in V} (f_1(u, v) + f_2(u, v)) = \sum_{v \in V} f_1(u, v) + \sum_{v \in V} f_2(u, v) = 0 + 0 = 0.$$

Hence flow conservation holds for $f_1 + f_2$.

tutors: Use this as an opportunity to point out the difference between this Q and the case when f_2 is a flow in the *residual network* (wrt f_1) - in that case everything has been set up for the capacity condition to also hold.

2. **Question:** we are given



Two questions:

- (a) Find a pair of subsets $X, Y \subseteq V$ such that $f(X, Y) = -f(V - X, Y)$.
- (b) Find a different pair of subsets $X, Y \subseteq V$ such that $f(X, Y) \neq -f(V - X, Y)$.

Answer: The point of this question is to get thinking about flow between *sets of vertices*, by applying Lemma 3 of Lecture slides 10-11. However, it might be good to think about specific examples of (a), (b) first, before looking at the details of what the pattern is.

What we are asking is: when is it the case that

$$f(X, Y) + f(V - X, Y) = 0?$$

Remember from Lemma 3 (part 3) of slides 10-11 that for any two *disjoint* sets $X', Y' \subset V$, and any other set Z' , and any flow f , we have $f(X', Z') + f(Y', Z') = f(X' \cup Y', Z')$. Observe that for our question, certainly X and $V - X$ are disjoint sets. Hence by Lemma 3 (3), we know

$$f(X, Y) + f(V - X, Y) = f(X \cup (V - X), Y) = f(V, Y).$$

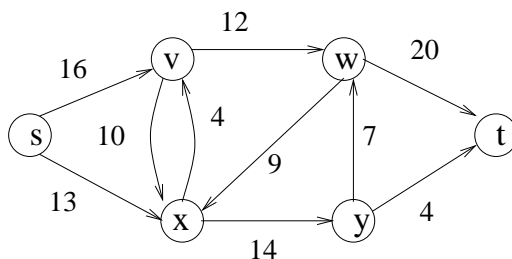
So we are testing whether $f(V, Y) = 0$ for (a), and whether $f(V, Y) \neq 0$ for (b) - once this is satisfied, X can be anything...

To make $f(V, Y) = 0$, we should either take Y such that $Y \cap \{s, t\} = \emptyset$, or $Y \cap \{s, t\} = \{s, t\}$. This can be seen by repeated application of part (3) of Lemma 3 from slides 10-11. To make $f(V, Y) \neq 0$, we should take Y such that $|Y \cap \{s, t\}| = 1$.

Here are some concrete examples of this behaviour:

- (a) As a concrete example, let $Y = \{v, x\}$. X can be *any* set, take $X = \{w\}$ as an example. Then $f(X, Y) = -12 + 4 = -8$. Then $f(V - X, Y) = 11 + 8 - 11 = 8$.
- (b) As a concrete example, take $Y = \{s\}$. Take $X = \{w\}$ again. Then we have $f(X, Y) = 0$. We have $f(V - X, Y) = -11 - 8 = -19$.

3. **Question:** execute the Ford-Fulkerson algorithm (*using the Edmonds-Karp heuristic*) on the Network below:

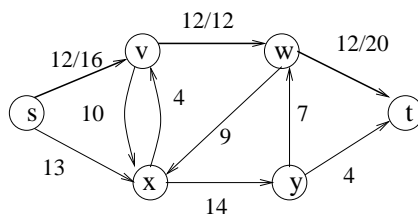


Answer: If we are using the Edmonds-Karp heuristic, then every time we search for an augmenting path, we must choose a shortest augmenting path.

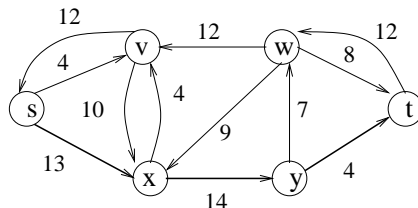
For our given network, we can see that on the first iteration, the path $p_1 = s \rightarrow v \rightarrow w \rightarrow t$ is a shortest path. We have $c(p_1) = 12$. Hence we define the flow $f_1 = f_{p_1}$ by

$$f_1(e) = f_{p_1}(e) = \begin{cases} 12 & \text{for } e = (s, v), (v, w), (w, t) \\ -12 & \text{for } e = (v, s), (w, v), (t, w) \\ 0 & \text{otherwise} \end{cases}$$

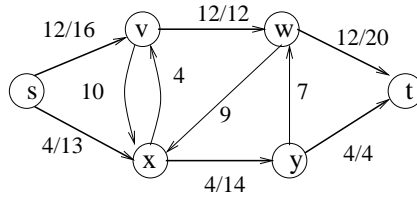
Pictorially, we have



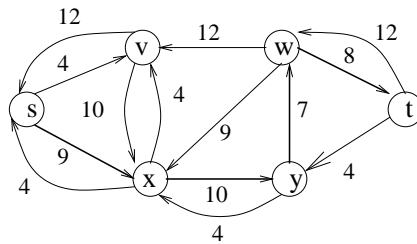
The *residual network* \mathcal{N}_{f_1} is as follows:



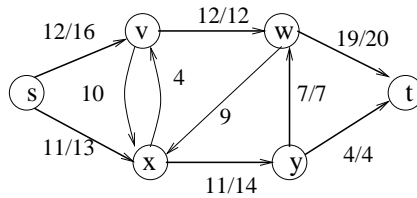
We now examine \mathcal{N}_{f_1} to find a shortest augmenting path. We find that $p_2 = s \rightarrow x \rightarrow y \rightarrow t$ is a shortest augmenting path in \mathcal{N}_{f_1} , min capacity 4, see above.... We therefore define a new flow f_{p_2} such that 4 units are shipped along the edges of the path p_2 , and -4 shipped in the backwards direction of p_2 . Then we define the flow $f_2 = f_1 + f_{p_2}$. Remember to point out this is possible *only* because f_1 is a flow in \mathcal{N} and f_2 is a flow in the *residual* network \mathcal{N}_{f_1} . Below is the flow $f_2 = f_1 + f_{p_2}$ in \mathcal{N} .



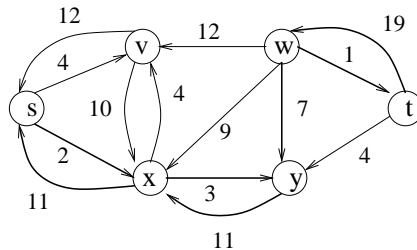
Below is the residual network \mathcal{N}_{f_2} . If we again try the Edmonds-Karp rule for finding an augmenting path of shortest possible length, we find the path $p_3 = s \rightarrow x \rightarrow y \rightarrow w \rightarrow t$ (this is of length 4, but there are no paths of length 3 or less in \mathcal{N}_{f_2}). The min capacity along the path is 4.



We define a new flow f_{p_3} in \mathcal{N}_{f_2} by shipping 7 units along p_3 . Then we define the flow f_3 in \mathcal{N} as $f_3 = f_2 + f_{p_3}$. The flow looks as follows:



We compute the residual network \mathcal{N}_{f_3} , see below for a picture.



By Ford-Fulkerson's algorithm, we now try for a (shortest) augmenting path in the \mathcal{N}_{f_3} . However, if we examine \mathcal{N}_{f_3} , we see that there is *no* augmenting path from s to t - the set of vertices accessible from s is now $\{s, v, x, y\}$.

Hence we terminate, returning the flow f_3 , of value 23.

4. **Question:** A well-known problem in graph theory is the problem of computing a *maximum matching* in a *bipartite graph* \mathcal{G} . Give an algorithm which shows how to solve this problem in terms of the network flow problem.

Definitions:

A (undirected) graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ is *bipartite* if we have $\mathcal{V} = \mathcal{L} \cup \mathcal{R}$ for two disjoint sets \mathcal{L}, \mathcal{R} , such that for every edge $e = (\mathbf{u}, \mathbf{v})$ exactly one of the vertices \mathbf{u}, \mathbf{v} lies in \mathcal{L} , and the other in \mathcal{R} .

A *matching* in an (undirected) graph G is a collection M of edges, $M \subseteq E$, such that for every vertex $v \in V$, v belongs to *at most one* edge of M .

A *maximum matching* is a matching of maximum cardinality (for a specific graph).

Answer:

To solve this question, we will design a network, based on the bipartite graph \mathcal{G} , where a maximum flow in the network corresponds to a maximum matching in \mathcal{G} .

Define the vertex set V' for our network \mathcal{N} to be $V' = \mathcal{L} \cup \mathcal{R} \cup \{s, t\}$, where s, t are two new distinguished vertices.

Define the (directed) edge set E' as follows:

$$E' = \{(s, \mathbf{u}) : \mathbf{u} \in \mathcal{L}\} \cup \{(\mathbf{u}, \mathbf{v}) : \mathbf{u} \in \mathcal{L}, \mathbf{v} \in \mathcal{R}, (\mathbf{u}, \mathbf{v}) \in E\} \cup \{(\mathbf{v}, t) : \mathbf{v} \in \mathcal{R}\}.$$

notice that the middle set in the union above is just the edge set E of the original graph, with all of these edges now directed from \mathcal{L} to \mathcal{R} .

Define the capacities of the network as follows:

$$\begin{aligned} c(s, \mathbf{u}) &= 1 && \text{for every } \mathbf{u} \in \mathcal{L} \\ c(\mathbf{u}, \mathbf{v}) &= 1 && \text{for every } \mathbf{u} \in \mathcal{L}, \mathbf{v} \in \mathcal{R}, (\mathbf{u}, \mathbf{v}) \in E \\ c(\mathbf{v}, t) &= 1 && \text{for every } \mathbf{v} \in \mathcal{R} \end{aligned}$$

I now claim that every flow of value k in \mathcal{N} corresponds to a matching of cardinality k in G . The max flow = maximum matching follows directly from this.

\Rightarrow Suppose f is a flow of value k in \mathcal{N} . We assume without any loss of generality that f is an integral flow (because all capacities are integers).

Recall that in \mathcal{N} , the vertex s has $|\mathcal{L}|$ neighboring edges (s, \mathbf{u}) . By definition of the value of a flow, $k = \sum_{\mathbf{u} \in \mathcal{V}} f(s, \mathbf{u}) = \sum_{\mathbf{u} \in \mathcal{L}} f(s, \mathbf{u})$. Therefore exactly k of the (s, \mathbf{u}) edges carry 1 unit of flow each (since no (s, \mathbf{u}) edge can carry more than 1).

Moreover by Lemma 11 in Lecture slides 13-14, every (S, T) cut in the network must be carrying flow of value k . Hence if we take $S = \{s\} \cup \mathcal{L}$, then we see there are exactly k (\mathbf{u}, \mathbf{v}) edges in the network which carry exactly 1 unit of flow from left to right (since no (\mathbf{u}, \mathbf{v}) edge can carry more than this).

Define $M = \{(\mathbf{u}, \mathbf{v}) \in E : f(\mathbf{u}, \mathbf{v}) = 1 \text{ in } \mathcal{N}\}$. Certainly $|M| = k$. I now show that M is a matching. For every $\mathbf{u} \in \mathcal{L}$, the flow conservation property must hold. For this

network, this means that for every $\mathbf{u} \in \mathbf{L}$, we require $(\sum_{\mathbf{v} \in \mathbf{R}} f(\mathbf{u}, \mathbf{v})) + f(\mathbf{u}, \mathbf{s}) = 0$. Therefore if $f(\mathbf{s}, \mathbf{u}) = 0$, we require $f(\mathbf{u}, \mathbf{v}) = 0$ for every $(\mathbf{u}, \mathbf{v}) \in \mathbf{E}$. If $f(\mathbf{s}, \mathbf{u}) = 1$ (so $f(\mathbf{u}, \mathbf{s}) = -1$), we require $f(\mathbf{u}, \mathbf{v}) = 1$ for exactly *one* $(\mathbf{u}, \mathbf{v}) \in \mathbf{E}$ (using our integer assumption). Hence every $\mathbf{u} \in \mathbf{L}$ will appear *at most once* in \mathbf{M} . We can use a similar argument to show that every $\mathbf{v} \in \mathbf{R}$ can appear at most once in \mathbf{M} . Hence \mathbf{M} is a matching.

\Leftarrow This is easier. Just explain how the matching of \mathbf{G} gets mapped to \mathcal{N} and check flow conservation.