

Advanced Algorithms and Data Structures

Hand-in 1

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The problem could be formulated and solved as a max-flow problem. There are a few things to observe before that: the network (as in the example) is undirected, in which case we would want to make it directed. We also have several sources that may themselves be connected (such as nodes 2 and 3 in the example).

Let the network be an undirected graph called $N = (V, E)$. We want to construct a max-flow graph $G = (V', E')$ from N . For all nodes $u, v \in V$ and $(u, v) \in E$, we add to our max-flow graph nodes u, v and an extra node w . We add directed edges $(u, v), (v, w)$ and (w, u) . If we let k denote the capacity of the edge $(u, v) \in E$, then we let $c(u, v) = c(v, w) = c(w, u) = k$ in G . Lastly, we add a super source s and an edge (s, v) to all the nodes v that were a source in N .

The number of edges in G is $|V'| = |V| + |E| + 1$ and $|E'| = 3|E| + p$, where p is the number of edges added from the super source s . p is exactly the number of sources in N .

Now we have the original problem stated as a max-flow problem, which can be easily expressed as a linear program

$$\begin{aligned} \max.: \quad & \sum_{v \in V_G} f(s, v) - \sum_{v \in V_G} f(v, s) \\ \text{s.t.:} \quad & f(u, v) \leq c(u, v), \quad \forall u, v \in V_G \\ & \sum_{v \in V_G} f(v, u) - \sum_{v \in V_G} f(u, v) = 0 \quad \forall u, v \in V_G \\ & f(u, v) = 0, \quad \forall u, v \in V_G \end{aligned}$$

As small example graph can be found in fig.1.

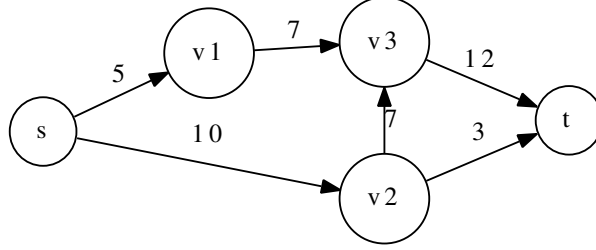


Figure 1: An example max-flow graph

The linear program corresponding to the graph in fig. 1 is the following:

$$\begin{array}{ll}
 \text{max.:} & f_{s,v_1} + f_{s,v_2} \\
 \text{s.t.:} & f_{s,v_1} \leq c(s, v_1) = 5 \\
 & f_{s,v_2} \leq c(s, v_2) = 10 \\
 & f_{v_1,v_3} \leq c(v_1, v_3) = 7 \\
 & f_{v_2,v_3} \leq c(v_2, v_3) = 7 \\
 & f_{v_2,t} \leq c(v_2, t) = 3 \\
 & f_{v_3,t} \leq c(v_3, t) = 12 \\
 & f_{s,v_1} = f_{v_1,v_3} \\
 & f_{s,v_2} = f_{v_2,v_3} + f_{v_2,t} \\
 & f_{v_1,v_3} + f_{v_2,v_3} = f_{v_3,t} \\
 & f_{s,v_1}, f_{s,v_2}, f_{v_1,v_3}, f_{v_2,v_3}, f_{v_2,t}, f_{v_3,t} \geq 0
 \end{array}$$

We have minimized the above linear program, such that edges with capacity zero have been omitted. Without showing the work of the simplex algorithm, we arrive at the solution: $f_{s,v_1} = 5, f_{s,v_2} = 10$.

In order to find the cheapest critical connection we have to identify the *minimal cuts*. Intuitively, if we have a cut (S, T) of our graph and a maximum flow f in that same graph, then the flow from S to T must be equal to capacity across that cut, $c(S, T)$. This means in order to increase the overall flow of the graph, we have to increase the capacities on some of the edges of our minimal cuts.

If we only have one such minimal cut, then we simply select the edge with the smallest capacity and increase its capacity. If we have more than one cut, it's not immediately clear which edges to choose. In our example above, we can identify three minimal cuts:

1. $S_1 = \{s\}, T_1 = V - \{s\}$
2. $S_2 = \{s, v_2\}, T_2 = \{v_1, v_3, t\}$

3. $S_3 = V - \{t\}$, $T_3 = \{t\}$.

In the first cut we can identify the edge (s, v_1) , $c(s, v_1) = 5$, as the cheapest to increase, for (S_2, T_2) the cheapest edge is (v_2, t) , $c(v_2, t) = 3$, and this is also the cheapest to increase for (S_3, T_3) . If we increase the capacities of these two edges by 1, then we have increased the capacities of all the minimal cuts (S_i, T_i) by 1. In this example the above cuts are still minimal cuts, so the max-flow min-cut theorem tells us that we have increased the maximum flow by 1 (otherwise we should have provoked another minimal cut).

Recaptured briefly, in order to find the “cheapest critical connection”, we need to inspect the minimal cuts and for each of them, select the edge with the smallest capacity to increase. In the end, by the max-flow min-cut property, we will have increased the maximum flow in the entire graph.

The dual of the above stated problem can be stated by restating the above program in standard form $Ax \leq b$ along with a coefficient vector $c = (1 \ 1 \ 0 \ 0 \ 0 \ 0)^T$. The dual is then given by $A^T y \geq c$, where we minimize $b^T y$. y is a vector of size proportional to the number of constraints, in our case 12, because each of the equality constraints give rise to two inequalities. Now we can state the dual:

$$\begin{array}{ll} \text{min.:} & 5y_1 + 10y_2 + 7y_3 + 7y_4 + 3y_5 + 12y_6 \\ \text{s.t.:} & \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & -1 & 1 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & -1 & 1 & 1 & -1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & -1 & 1 \end{bmatrix} y \geq \begin{pmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \end{array}$$

A solution to the dual problem is $y_1 = 1, y_2 = 1$ and the rest are zero.

In order to model that some of the relay stations aren't owned by the company, we can rephrase the problem as a *minimum-cost flow* problem (Cormen 29.51, pg 862), where the cost function is defined by:

$$a(u, v) = \begin{cases} 1, & \text{if } u \text{ xor } v \text{ is not owned by the company} \\ 0, & \text{otherwise.} \end{cases}$$

Then we can directly apply the definition from the book, because we can require the value of the flow to be exactly d .

Results

Obtaining results

Using the provided *Network.java*, the following steps were taken to obtain the results:

1. The given network was converted to a maximum flow graph
2. Then it was further transformed into a linear program: As described in Corman, pg 860, an objective function and constraints for the capacities and the flow conservation have been generated.
3. The objective function was set to be maximized.
4. This LP has been solved using *lp_solve* with the Java wrapper.

The returned result was the maximum flow and represented the largest bit rate. For solving the problem of the lowest leasing rate, the linear program has been modified to a minimum cost flow problem (see Cormen pg 861):

1. The previous objective function was used as the LHS of a new constraint, the previous result was used as the RHS.
2. A cost matrix was created and transformed into a form to match the converted network.
3. A new objective function using the cost matrix was generated.
4. The new LP minimized the objective function.

This returned the minimum costs for leasing while maintaining the transfer rate.

See attached Java files, especially *MF.java* and the modified *Network.java*, also see a simpler network which represents the example from above in *Small-Net.java*.

The following is the relevant part of the output of the calculations on the given network.

Largest bit rate

Max Flow: 95.0	Flow[(1[s])(8)] = 10.0
Flow[(1[s])(0[s])] = 30.0	Flow[(2[s])(9)] = 25.0
Flow[(1[s])(1[s])] = 10.0	Flow[(3[s])(9)] = 5.0
Flow[(1[s])(2[s])] = 25.0	Flow[(4[s])(9)] = 5.0
Flow[(1[s])(3[s])] = 5.0	Flow[(4[s])(10)] = 10.0
Flow[(1[s])(4[s])] = 15.0	Flow[(5[s])(10)] = 10.0
Flow[(1[s])(5[s])] = 10.0	Flow[(6)(18)] = 20.0
Flow[(0[s])(6)] = 20.0	Flow[(7)(18)] = 10.0
Flow[(0[s])(7)] = 10.0	Flow[(8)(12)] = 10.0

Flow[(9)(11)] = 15.0
 Flow[(9)(12)] = 20.0
 Flow[(10)(11)] = 20.0
 Flow[(11)(13)] = 15.0
 Flow[(11)(14)] = 20.0
 Flow[(12)(15)] = 30.0
 Flow[(13)(14)] = 10.0
 Flow[(13)(15)] = 5.0
 Flow[(14)(19[t])] = 30.0
 Flow[(15)(16)] = 10.0
 Flow[(15)(19[t])] = 30.0
 Flow[(16)(19[t])] = 35.0
 Flow[(17)(23)] = 10.0
 Flow[(18)(20)] = 10.0

Flow[(18)(21)] = 10.0
 Flow[(18)(18/17)] = 10.0
 Flow[(20)(22)] = 10.0
 Flow[(21)(22)] = 20.0
 Flow[(22)(24)] = 30.0
 Flow[(23)(23/21)] = 10.0
 Flow[(24)(24/15)] = 5.0
 Flow[(24)(24/16)] = 25.0
 Flow[(24/15)(15)] = 5.0
 Flow[(24/16)(16)] = 25.0
 Flow[(18/17)(17)] = 10.0
 Flow[(23/21)(21)] = 10.0

Lowest leasing price

Minimum Cost: 60.0
 Flow[(S)(0[s])] = 30.0
 Flow[(S)(1[s])] = 10.0
 Flow[(S)(2[s])] = 25.0
 Flow[(S)(3[s])] = 5.0
 Flow[(S)(4[s])] = 15.0
 Flow[(S)(5[s])] = 10.0
 Flow[(0[s])(6)] = 20.0
 Flow[(0[s])(7)] = 10.0
 Flow[(1[s])(8)] = 10.0
 Flow[(2[s])(9)] = 25.0
 Flow[(3[s])(9)] = 5.0
 Flow[(4[s])(9)] = 5.0
 Flow[(4[s])(10)] = 10.0
 Flow[(5[s])(10)] = 10.0
 Flow[(6)(18)] = 20.0
 Flow[(7)(18)] = 10.0
 Flow[(8)(12)] = 10.0
 Flow[(9)(11)] = 15.0
 Flow[(9)(12)] = 20.0
 Flow[(10)(11)] = 20.0

Flow[(11)(13)] = 15.0
 Flow[(11)(14)] = 20.0
 Flow[(12)(15)] = 30.0
 Flow[(13)(14)] = 5.0
 Flow[(13)(15)] = 10.0
 Flow[(14)(19[t])] = 25.0
 Flow[(15)(16)] = 10.0
 Flow[(15)(19[t])] = 30.0
 Flow[(16)(19[t])] = 40.0
 Flow[(17)(23)] = 10.0
 Flow[(18)(20)] = 10.0
 Flow[(18)(21)] = 10.0
 Flow[(18)(18/17)] = 10.0
 Flow[(20)(22)] = 10.0
 Flow[(21)(22)] = 10.0
 Flow[(22)(24)] = 20.0
 Flow[(23)(24)] = 10.0
 Flow[(24)(24/16)] = 30.0
 Flow[(24/16)(16)] = 30.0
 Flow[(18/17)(17)] = 10.0