

Chapter 14, Network flow problems

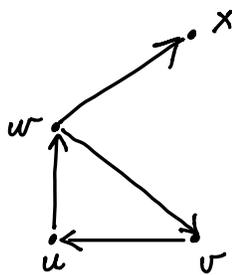
Mathematical models for flows in networks.

Models will be linear, and based on graphs ^{network}

A directed graph is an ordered pair (\mathcal{N}, A) where

- \mathcal{N} is a finite set, called nodes (vertices, or points)
- A is a finite set of ordered pairs of nodes, called edges (lines, or arcs)

Drawing a directed graph: (i, j) denotes the edge from node i to node j (use arrows)



i : initial node } end nodes of $e = (i, j)$
 j : terminal node }

Here:

$$\mathcal{N} = \{u, v, w, x\}$$

$$A = \{(v, w), (u, w), (w, v), (w, x)\}$$

A flow in a directed graph:

- Goods flow along the edges in the direction the edges show
- Each node has a certain supply or demand for goods,

denoted b_i :

a) $b_i > 0$: supply node

b) $b_i < 0$: demand node

c) $b_i = 0$: transshipment node

We will have that $\sum_{i \in \mathcal{N}} b_i = 0$: Everything stays within the graph:
produced in graph \Rightarrow consumed in graph.

x_{ij} denotes the flow in graph in edge (i, j)

X_{ij} often demanded to be nonnegative, and with an upper capacity: $0 \leq X_{ij} \leq \Delta_{ij}$
not used for now

With shipping along (i,j) there is associated a cost C_{ij} (the cost of shipping one unit). Total cost of shipping in graph is

$$\sum_{(i,j) \in A} C_{ij} X_{ij}$$

Flow balance equations:

$$\underbrace{\sum_{i: (i,k) \in A} X_{i,k}}_{\text{inflow to } k} - \underbrace{\sum_{j: (k,j) \in A} X_{k,j}}_{\text{outflow to } k} = -b_k$$

$b_k > 0$: supply node:
 outflow > inflow
 inflow - outflow < 0

multiply with -1 on both sides:

$$\text{outflow} - \text{inflow} = b_k \quad (\text{easier to remember}).$$

Minimum cost network flow problem (MCF)

$$\begin{aligned} \min \quad & \sum_{(i,j) \in A} C_{ij} X_{ij} \\ \text{s.t.} \quad & \sum_{(i,k) \in A} X_{i,k} - \sum_{(k,j) \in A} X_{k,j} = -b_k \\ & x \geq 0 \end{aligned}$$

This is an LP problem with a special structure.

In matrix form:

$$(*) \quad \begin{aligned} \min \quad & C^T X \\ \text{s.t.} \quad & AX = -b \\ & X \geq 0 \end{aligned}$$

A : one row for each node ($|N|$ rows)
 one column for each edge (i,j) .

In column (i,j) , entry j is 1 , entry i is -1 ,
 all other entries are zero.

in a row: a 1 for each incoming edge, and
 a -1 for each outgoing edge.

A is also called a node-edge incidence matrix

Dual problem of $(*)$ is

$$\begin{array}{l}
 (**) \quad \max -b^T y \\
 \text{s.t.} \quad A^T y + z = c \\
 \quad \quad z \geq 0
 \end{array}$$

equality constraints in $(*)$
 \Downarrow
 y is unconstrained

y : one component for each node

z : one component for each edge

In component form:

$$\begin{array}{l}
 \max -\sum b_i y_i \\
 \text{s.t.} \quad y_j - y_i + z_{ij} = c_{ij} \quad (i,j) \in A \\
 \quad \quad z_{uv} \geq 0 \quad y_i \text{ unconstrained}
 \end{array}$$

Equivalent to:

$$\begin{array}{l}
 -\min \sum b_i y_i \\
 \text{s.t.} \quad y_j - y_i \leq c_{ij} \quad (i,j) \in A
 \end{array}$$

This problem is feasible (set $y_i = 0$).

LP theory to find optimal solution to MCF:

$$\left. \begin{array}{l} \text{complementary slack: } X_{uv} Z_{uv} = 0 + \text{feasibility} \\ \text{(slacks in primal are 0)} \end{array} \right\} \begin{array}{l} \text{primal: } X \geq 0 \\ \text{flow balance} \\ \text{dual: } y_j - y_i \leq C_{ij} \\ \text{all } (i,j) \end{array}$$

\Downarrow
if $y_j < y_i + C_{ij}$ ($\Leftrightarrow Z_{ij} > 0$), then X_{ij} must be zero

MCF can be solved far more efficiently than general LP problems, due to their special structure.

Terms from graph theory

A path: a sequence v_1, v_2, \dots, v_k of nodes where either (v_i, v_{i+1}) or (v_{i+1}, v_i) is in A for each i . v_1 and v_k are called end nodes of the path.

Connected graph: A graph where there is a path between any pair of nodes.

Cycle: A path where the end nodes coincide

A subgraph of (N, A) : A network (N', A') where $N' \subseteq N$, $A' \subseteq A$, and all edges in A' have end nodes in N' .

A tree: A connected graph without cycles.
(the number of edges = number of nodes - 1)



Spanning tree: A subgraph (N, A') without cycles.

We will explain that:

1) spanning trees correspond to LP bases

2) Calculation of basic solutions: performed by simple operations on spanning trees.

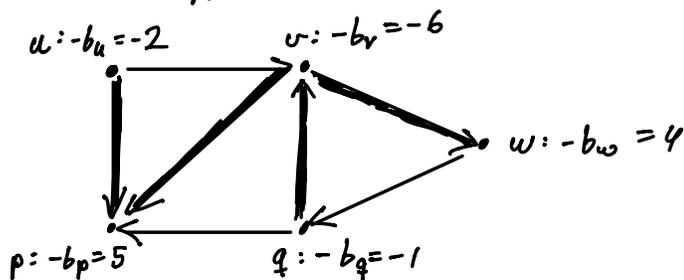
3) pivot \rightarrow simple transformation of a spanning tree into another spanning tree.

For a given spanning tree we compute a so-called tree solution X as follows:

1) Set $X_e = 0$ for all $e = (i, j)$ not in the spanning tree.
(X_{ij} nonbasic)

2) Starting at a leaf node, we can use the flow balance equation to compute each X_{ij} , where (i, j) is in the spanning tree (X_{ij} basic)

Example: spanning tree in bold
supplies/demands shown at the nodes.



1. $X_{uv} = X_{qp} = X_{wq} = 0$ (these are not in the spanning tree)

Four edge flows need to be computed. 3 leaf nodes:

2. u is leaf node: outflow - inflow = $b_u = 2$

$$\underline{X_{up}} = 2$$

3. w is leaf node: outflow - inflow = $b_w = -4$

$$- X_{vw} = -4$$

$$\underline{X_{vw}} = 4$$

4. q is leaf node: outflow - inflow = $b_q = 1$

$$\underline{X_{qv}} = 1$$

5. Flow balance in p : outflow - inflow = $b_p = -5$

$$-X_{up} - X_{vp} = -5$$

$$-2 - X_{vp} = -5$$

$$\underline{X_{vp} = 3}$$

Alternatively:

Flow balance in v : outflow - inflow = $b_v = 6$

$$X_{vp} + X_{wv} - X_{qv} = 6$$

$$X_{vp} + 4 - 1 = 6$$

$$\underline{X_{vp} = 3}$$

Algorithm for computing x given the spanning tree T :

1) Choose a leaf node in the spanning tree

2) Apply flow balance for that node to compute x_e for the unique edge in the spanning tree incident to that leaf.

3) Remove e (and leaf node) from T , and continue from 1) (x_e is needed in the other flow balance equations).

This procedure will always work, to find a tree solution.

Note: One flow balance equation is not used.

The equations we solved (variables are permuted)

$$2. \quad X_{up} = 2$$

$$3. \quad -X_{wv} = -4$$

$$4. \quad X_{qv} = 1$$

$$5. \quad -X_{up} - X_{vp} = -5$$

alt. way: $X_{wv} - X_{qv} + X_{vp} = 6$

See that this gives a 4×4 invertible submatrix of A (since the matrix is lower triangular)

$$B = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -1 & 0 & 0 & -1 \end{pmatrix}$$

Each row: Flow balance for $|N|-1$ nodes (only one can be dropped)

Each column: edges in the spanning tree

Convince yourself: Any spanning tree gives rise to a $(|N|-1) \times (|N|-1)$ submatrix which is invertible. From this we can find x_e for e in the spanning tree. The others are set to 0.

If a graph is connected, we always have a spanning tree.

A spanning tree will always give linearly independent rows as above, so $\text{rank } A \geq |N|-1$

Also $\text{rank}(A) < N$, since the sum of all rows is 0.

It follows that $\text{rank } A = |N|-1$

So: Given a spanning tree, the corresponding tree solution can be found as the basic solution where any node is taken out of the flow balance equations called root node

Basic variables: Edges in the spanning tree

Nonbasic variables: Edges outside the spanning tree.

Removing root node, the problem looks as follows:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & \tilde{A} x = -\tilde{b} \\ & x \geq 0 \end{aligned}$$

A with one row removed

Theorem 14.1: A square submatrix of \tilde{A} gives a basis if and only if the corresponding edges form a spanning tree.

Proof: From the above example: Any spanning tree gives a basis (since the matrix we end up with is triangular). Assume now that the columns do not form a spanning tree. Then we have $|N|-1$ edges that do not visit all $|N|$ nodes. Graph theory says that we then must have a cycle. Corresponding columns sum to 0, so we have linear dependence of the columns, so the matrix is not invertible \Rightarrow no basis. \blacksquare

So: spanning trees \Leftrightarrow LP bases in \tilde{A}

How do we find the dual variables?

$$\max - \sum_{i \in N} b_i y_i$$

$$y_j - y_i + z_{ij} = c_{ij} \quad (i,j) \in A$$

Dual variable for the root node (y_r).
is equivalent to having $y_r = 0$.

if (i,j) in spanning tree: X_{ij} is basic $\Rightarrow Z_{ij}$ is nonbasic $\Rightarrow z_{ij} = 0$

$$y_j - y_i + z_{ij} = c_{ij}$$

$$y_j - y_i = c_{ij}$$

All y_i can be found from this (use that $y_r = 0$)

After the y_i have been computed, we can use \leftarrow to compute the z_{ij} .