

**UNCLASSIFIED**

**AD 4 2 2 8 4 2**

**DEFENSE DOCUMENTATION CENTER**

**FOR**

**SCIENTIFIC AND TECHNICAL INFORMATION**

**CAMERON STATION, ALEXANDRIA, VIRGINIA**



**UNCLASSIFIED**

NOTICE: When government or other drawings, specifications or other data are used for any purpose other than in connection with a definitely related government procurement operation, the U. S. Government thereby incurs no responsibility, nor any obligation whatsoever; and the fact that the Government may have formulated, furnished, or in any way supplied the said drawings, specifications, or other data is not to be regarded by implication or otherwise as in any manner licensing the holder or any other person or corporation, or conveying any rights or permission to manufacture, use or sell any patented invention that may in any way be related thereto.

⑤ 739200

①

AD No. 422842  
DDC FILE COPY



\$1.60

DDC  
NOV 15 1963  
RECEIVED  
TISIA A

5 739200

6 NETWORK FLOW THEORY,  
10 by L. R. Ford, Jr.  
P-923 ✓  
August 14, 1956

B

The RAND Corporation  
1700 MAIN ST. • SANTA MONICA • CALIFORNIA

SUMMARY

↓  
The labeling algorithm for the solution of maximal network flow problems and its application to various problems of the transportation type are discussed. ↑

## NETWORK FLOW THEORY

L. R. Ford, Jr.

1. INTRODUCTION

A network (or linear graph) is a collection of points or nodes, some of which may be joined together by arcs. We shall denote the points by  $P_i$ ,  $i = 0, 1, 2, \dots, N$ , and denote the arc joining  $P_i$  to  $P_j$  in that order by  $A_{ij}$ . (Note that there may also be an arc  $A_{ji}$  joining  $P_j$  to  $P_i$ .) We may also have associated with the arc  $A_{ij}$  a capacity  $c_{ij}$  and a length (or cost)  $l_{ij}$ . We shall assume these to be positive integers. We shall also distinguish two points in the network,  $P_0$ , the origin, and  $P_N$ , the terminal. One may think of this system as a rail network, in which the origin represents a factory or a warehouse at which goods enter the system, and the terminal represents a consumer for these goods who removes them from the system. The  $c_{ij}$  then represent upper bounds on the shipping capacity from  $P_i$  to  $P_j$  and the  $l_{ij}$  may represent variously the distance from  $P_i$  to  $P_j$  or the time required to ship from  $P_i$  to  $P_j$  or the unit cost of shipping from  $P_i$  to  $P_j$ . (We shall agree that if  $c_{ij} = 0$  then  $l_{ij} = \infty$  and that  $l_{ij} \neq 0$ .)

Evidently many different problems may be posed with this framework. We shall discuss primarily the following three problems, stated verbally below.

- A. To find a maximal steady state flow of goods from the origin to the terminal, independent of cost considerations (i.e. of  $l_{ij}$ ).

- B. To find the cheapest route from origin to terminal independent of capacity constraints (i.e. of  $c_{ij}$ ).
- C. To find the maximal amount of goods that can be shipped from origin to terminal in a given time,  $T$  (here interpreting  $l_{ij}$  as the time required to ship from  $P_i$  to  $P_j$ ).

We shall discuss A first as being by far the richest in applications. It appears as a subproblem in a very large number of transportation-type problems, and from the theory which we shall develop one may obtain a large number of combinatorial results as well.

## 2. FORMAL STATEMENT OF PROBLEM A

A steady state flow of goods from origin to terminal is required to have the following properties:

- (a) the amount flowing into  $P_i$  must equal the amount flowing out of  $P_i$  for  $i \neq 0, N$ ;
- (b) the amount flowing along  $A_{ij}$  must be less than or equal to  $c_{ij}$ .

Subject to these restrictions, it is desired to maximize  $x_P$ , the amount flowing out of  $P_0$ , or equivalently, the amount flowing into  $P_N$ . This may be presented as a linear programming problem as follows. Denote by  $x_{ij} \geq 0$  the flow from  $P_i$  to  $P_j$ . Then

$$(1) \left\{ \begin{array}{l} \sum_j x_{1j} - \sum_j x_{j1} = 0 \quad (i = 1, 2, \dots, N-1) \\ \sum_j x_{0j} - \sum_j x_{j0} = x_F \\ \sum_j x_{Nj} - \sum_j x_{jN} = -x_F \\ 0 \leq x_{1j} \leq c_{1j} \\ \text{maximize } x_F . \end{array} \right.$$

Thus Problem A could be solved by a direct application of the simplex method to the above system.

### 3. SOLUTION OF PROBLEM A BY LABELING

We here outline an iterative technique which will solve Problem A, starting with any flow whatsoever, and will also solve the dual problem to be discussed later. The labeling process attaches labels of the form  $(P_{j,h}^+)$  to certain points in the following manner.

(a) Label  $P_0$  with the label  $(-, \infty)$ .

(b) Take any labeled point  $P_j$  not yet scanned.

Suppose labeled  $(P_k^+, h)$ . To all points  $P_i$  which are unlabeled and such that  $c_{ji} - x_{ji} > 0$  attach the label  $(P_j^+, \min [h, c_{ji} - x_{ji}])$ . To all points  $P_i$  which are now unlabeled, and such that  $x_{ij} > 0$ , attach the label  $(P_j^-, \min [h, x_{ij}])$ . [In case  $P_i$



is a candidate for a label in several ways, use any applicable label.]

If the terminal  $P_N$  is ever labeled the process ceases immediately. Otherwise the process continues until no new labels are obtained, and the terminal is left unlabeled.

In the former case we increase the flow as follows. If  $P_N$  is labeled  $(P_k^+, h)$  we replace  $x_{kN}$  by  $x_{kN} + h$ , and if  $P_N$  is labeled  $(P_k^-, h)$  we replace  $x_{Nk}$  by  $x_{Nk} - h$ . In either case we then turn our attention to  $P_k$ . In general, if  $P_k$  is labeled  $(P_j^+, m)$  replace  $x_{jk}$  by  $x_{jk} + h$ , and if labeled  $(P_j^-, m)$  replace  $x_{kj}$  by  $x_{kj} - h$ , in either case turning attention then to  $P_j$ . Eventually we arrive back at the origin and the process ceases, having increased  $x_F$  by  $h$  units. Then all labels are erased and the labeling process is begun again. We see easily that equations (1) remain satisfied.

In the latter case we are done. We shall defer proof of this momentarily until after we discuss cuts.

Definition: A cut in a network is any set of arcs whose removal disconnects the origin from the terminal. The value of a cut is the sum of the capacities of its arcs.

It is clear that, given a cut, any flow from origin to terminal must pass through it. Thus  $x_F \leq$  value of any cut.

We shall prove that we have solved Problem A by producing a cut whose value is equal to  $x_F$ . Note that we have finally achieved a labeling which includes the origin and excludes the

terminal. If we let  $I$  be the set of indices of labeled points and  $J$  the set of indices of unlabeled points, then the arcs  $A_{ij}$  for  $i \in I, j \in J$  evidently form a cut. For these arcs  $x_{ij} = c_{ij}$  and  $x_{ji} = 0$ , for otherwise  $P_j$  would be labeled. Summing equations (1) over  $i \in I$  only yields

$$x_F = \sum_{\substack{i \in I \\ j \in J}} x_{ij} = \sum_{\substack{i \in I \\ j \in J}} c_{ij} = \text{value of the cut.}$$

Hence  $x_F$  is maximal; also the cut is one of minimal value.

This, in passing, is a constructive proof of the minimal cut theorem for the case of integral (or rational)  $c_{ij}$ .

The Minimal Cut Theorem. In any network the maximal steady state flow is equal to the minimal cut value.

#### 4. A SET OF DUAL VARIABLES

Writing the last inequalities of (1) in the form  $x_{ij} + y_{ij} = c_{ij}$ , we find the dual problem: To determine  $\pi_i$  (point prices) and  $y_{ij}$  (arc prices) satisfying

$$(2) \quad \left\{ \begin{array}{l} \pi_i - \pi_j + y_{ij} \geq 0 \\ y_{ij} \geq 0 \\ \pi_N - \pi_0 \geq 1 \\ \text{minimize } \sum c_{ij} y_{ij} . \end{array} \right.$$

Letting  $\pi_i = 0$  for  $i \in I$ ,  $\pi_i = 1$  for  $i \in J$  and  $\gamma_{ij} = 1$  for  $i \in I$ ,  $j \in J$ ,  $\gamma_{ij} = 0$  otherwise, yields a set of prices which satisfy (2) and which produce a dual form equal to the primal form (via the Minimal Cut Theorem). Hence these prices form an optimal solution to the dual system, from the duality theorem.

##### 5. CERTAIN PROBLEMS SOLVABLE BY NETWORK FLOW METHODS

There are certain transportation-type problems which may be viewed as maximal flow problems, and others in which such a problem appears as an auxiliary or an associated problem.

( $\alpha$ ) The Capacitated Hitchcock Problem. The problem may be stated as follows. To find  $x_{ij} \geq 0$  such that

$$\sum_j x_{ij} = a_i \quad \sum_i x_{ij} = b_j$$

$$0 \leq x_{ij} \leq c_{ij}$$

$$\text{minimize } \sum a_{ij} x_{ij} .$$

Clearly, for a solution to exist,  $\sum a_i$  must equal  $\sum b_j$ . But even if this requirement is satisfied the  $c_{ij}$  may make a solution impossible.

The question of feasibility may be settled, however, by the flow algorithm. Set up the network with points  $O$ ,  $A_i$ ,  $B_j$ , and  $\mathcal{T}$  where  $O$  is the origin and  $\mathcal{T}$  the terminal. The arcs in the network are

$\widehat{OA}_1$  with capacity  $a_1$ ,  $\widehat{B_jT}$  with capacity  $b_j$ , and  $\widehat{A_1B_j}$  with capacity  $c_{1j}$ . The maximal flow in this network is  $\sum a_1$  if and only if the original problem is feasible.

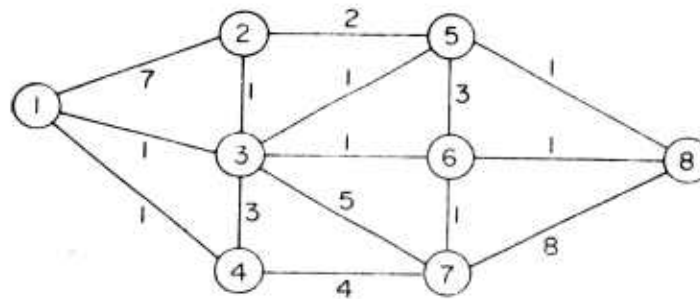
- (β) The Capacitated Transshipment Problem. Here the problem is in a capacitated network; certain points,  $O_1$ , have availabilities of  $a_1$  and other points,  $T_j$ , have requirements  $b_j$ . The feasibility of this problem may be settled by adding a new origin  $O$  joined to  $O_1$  by an arc of capacity  $a_1$ , and a new terminal  $T$  with  $T_j$  joined to  $T$  by an arc of capacity  $b_j$ . Again, if the maximal flow is  $\sum a_1 = \sum b_j$ , then the system is feasible.

6. TWO EXAMPLES

- (α) Check the feasibility of this Capacitated Hitchcock problem.

		$\overbrace{\hspace{10em}}^{b_j}$				
		9	10	9	12	
$a_1$	$\left\{ \begin{array}{l} 13 \\ 10 \\ 7 \\ 10 \end{array} \right.$	3	4	2	4	$= c_{ij}$
		1	3	2	4	
		5	4	3	2	
		3	3	2	1	

(β) Check the feasibility of this Capacitated Trans-shipment Problem,



where the availabilities are 3 units at  $P_1$  and 5 units at  $P_6$ ; the requirements are 4 units at  $P_4$  and 4 units at  $P_8$ .

7. A SOLUTION FOR PROBLEM B

This may be set up as a linear programming problem by starting with one unit available at the origin, one unit demanded at the terminal, and requiring the minimization of distance traveled. The equations are

$$(3) \quad \begin{cases} \sum (x_{0j} - x_{j0}) = 1 \\ \sum (x_{ij} - x_{ji}) = 0 \\ \sum (x_{Nj} - x_{jN}) = -1 \\ \text{minimize } \sum l_{ij} x_{ij} \end{cases}$$

The dual problem, with which we shall work is

$$(3') \quad \begin{cases} x_i + l_{ij} \geq x_j \\ x_0 = 0 \\ \text{maximize } x_n \end{cases}$$

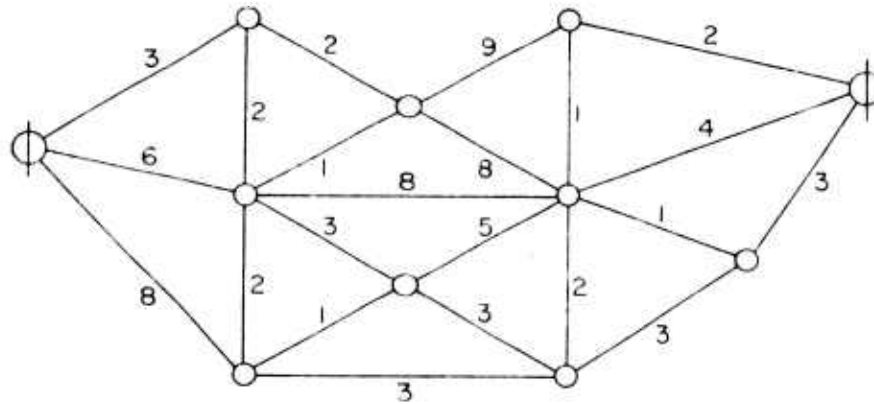
where the  $x_i$  are dual variables associated with the points of the graph and, at least in the optimal solution,  $x_i$  may be thought of as representing the shortest distance from  $P_0$  to  $P_i$ .

A computing procedure is the following. Assign initially  $x_0 = 0$  and  $x_i = \infty$  for  $i \neq 0$ . Scan the network for a pair  $P_i$  and  $P_j$  with the property that  $x_i - x_j > l_{ji}$ . For this pair replace  $x_i$  by  $x_j + l_{ji}$ . Continue this process. Eventually no such pairs can be found, and  $x_N$  is now minimal and represents the minimal distance from  $P_0$  to  $P_N$ . Clearly, if no such pairs can be found, the system (3) is satisfied. We shall now prove optimality.

Let  $P_0, P_{i_1}, \dots, P_{i_m}, P_N$  be the shortest path. Along this path,  $x_{i_{k+1}} - x_{i_k} \leq l_{i_k i_{k+1}}$ . Summing these relations gives  $x_N \leq$  (length of shortest path). On the other hand, for each  $P_j$  ( $j \neq 0$ ) there is some  $P_i$  with  $x_i + l_{ij} = x_j$ ,  $l_{ij} \neq 0$ . For if  $j \neq 0$ ,  $x_j$  started at  $\infty$  and has been decreasing monotonely. If  $x_j$  is still  $\infty$ , we are done; if not, at the last decrease there was such an  $i$  which must still be at the same value. Tracing out this chain backwards from  $P_N$  the  $x_i$  are monotone strictly decreasing; eventually the origin  $P_0$  is reached. Here  $x_j - x_i = l_{ij}$  and summing these gives  $x_N =$  length of this chain. Hence this is the shortest path.

8. AN EXAMPLE

Find the shortest path through the following network.



9. PROBLEM C VIEWED AS PROBLEM A

Problem C, the dynamic network flow problem, may be viewed as a static flow in the following larger network. Let  $S$  be the network whose nodes are points  $P_i^\tau, P_j^\tau$  corresponding to the old  $P_i$  but one for each basic time unit, i.e., for all  $\tau$  with  $0 \leq \tau \leq T$ . We connect these points with directed capacitated arcs as follows.

$$\left[ \begin{array}{l} \text{Arc } \widehat{P_i^\tau P_i^{\tau+1}} \text{ is present with infinite capacity for} \\ 0 \leq \tau \leq T-1 \text{ and all } i. \\ \text{Arc } P_i^\tau P_j^{\tau+l_{ij}} \text{ is present with capacity } c_{ij} \text{ for} \\ 0 \leq \tau \leq T - l_{ij} \text{ and all } i, j. \end{array} \right.$$

One sees fairly readily that a maximal steady state flow in  $S$  is a maximal dynamic flow for  $T$  time periods in the original network.

## 10. THE TANKER SCHEDULING PROBLEM

This problem may be stated as follows. We have given  $m$  pickup points  $P_1$  and  $n$  discharge points  $Q_j$ , and a schedule  $\{t_{1j}^k\}$  where  $t_{1j}^k$  is the time at which a tanker is required to be at  $P_1$  to pick up a load destined for  $Q_j$ . (There may be several such times; these are distinguished by the index  $k$ .) We have further given two arrays of positive numbers,  $a_{1j}$  and  $b_{1j}$ , where  $a_{1j}$  represents the loading-traveling time from  $P_1$  to  $Q_j$  and  $b_{1j}$  the unloading-traveling time from  $Q_j$  to  $P_1$ . The problem is to meet the schedule with a minimal number of tankers.

This problem has been solved as a transportation problem by Dantzig and Fulkerson in [2]. Here it will be shown to be essentially of type A above.

Let  $T$  be the maximal  $t_{1j}^k$  and set up a network with points  $P_1^\tau, Q_j^\tau$  for  $0 \leq \tau \leq T$ . We join  $Q_j^\tau$  to  $P_1^{\tau+a_{1j}}$  with an arc of infinite capacity,  $0 \leq \tau \leq T - a_{1j}$ , and  $P_1^\tau$  to  $P_1^{\tau+1}, Q_j^\tau$  to  $Q_j^{\tau+1}$  with an arc of infinite capacity,  $0 \leq \tau \leq T-1$ . The schedule may be interpreted as providing a requirement of one unit at  $P_1^{t_{1j}^k}$  and an availability of one unit at  $Q_j^{t_{1j}^k+a_{1j}}$  for each  $t_{1j}^k$  [such that  $t_{1j}^k \leq T - a_{1j}$ ]. This can be accomplished by joining each such  $P_1^{t_{1j}^k}$  to a terminal,  $\mathcal{J}$ , by an arc of capacity one, and each such  $Q_j^{t_{1j}^k+a_{1j}}$  to an origin,  $O$ , by an arc of capacity one.

Evidently the maximal flow in this network represents the maximal reassignment potential available in the system.



Note that this problem is not, strictly speaking, of type C, since the availabilities are not the same in each time period. A new algorithm to handle type C directly, without turning to the dynamic network, has recently been developed by D. R. Fulkerson and the author.

#### BIBLIOGRAPHY

1. Dantzig, G. B., and D. R. Fulkerson, On the Min Cut Max Flow Theorem of Networks, The RAND Corporation, Research Memorandum RM-1418, Jan. 1, 1955 (to appear in Contributions to Linear Inequalities and Related Topics, Annals of Math. Study No. 38).
2. Dantzig, G. B., and D. R. Fulkerson, "Minimizing the Number of Tankers to Meet a Fixed Schedule," Naval Research Logistics Quarterly, Vol. 1, No. 3., Sept. 1954.
3. Ford, L. R., Jr., and D. R. Fulkerson, Maximal Flow Through a Network, The RAND Corporation, Paper P-605, Nov. 19, 1954 (to appear in the Canadian Journal of Math.).
4. Ford, L. R., Jr., and D. R. Fulkerson, A Simple Algorithm for Finding Maximal Network Flows and an Application to the Hitchcock Problem, The RAND Corporation, Paper P-743, September 26, 1955.
5. Ford, L. R., Jr., and D. R. Fulkerson, A Primal-Dual Algorithm for the Capacitated Hitchcock Problem, The RAND Corporation, Paper P-827, March 23, 1956.
6. Gale, David, A Theorem on Flows in Networks, The RAND Corporation, Paper P-798, January 3, 1956 (to appear in the Pacific Journal of Math.).