

# 最短路問題の“最適”算法について

東京大学 工学部 中森真理雄  
伊理正夫

"Optimal" Algorithms for the All-Shortest-Path  
Problem on a Network

By

Mario NAKAMORI and Masao IRI

Faculty of Engineering, University of Tokyo

## *Abstract*

A class of all-shortest-path algorithms expressed as repeated applications of triple-operations is considered. Triple-operations are regarded as operators which induce transformations on distance matrices, and the algebraic structure of the operator semigroup is investigated from the points of view of equivalence relation, partial order, etc. On such investigation, some of the existing algorithms for a complete network are shown to be optimal, and useful formulae for the formal manipulation of algorithms are listed, by means of which the validity of various algorithms is proved. Also, some specially structured networks are discussed and new optimal algorithms for them are proposed.

## 1. *Introduction*

The all-shortest-path problems on a network are not only of great significance by themselves but also play important rôles as subproblems of many larger ones of network-flow type in operations research. There have been proposed for this kind of problems many methods of solution, "new" proposals still continuing at present.

Expreience shows that many efficient methods belong to the family of those algorithms which are expressed as repeated applications of the so-called triple-operations (or, operations of min-addition pivoting) on distance matrices. In the present paper we aim at establishing a systematic algebraic theory by which to unify the algorithms in this family.

In §2, triple-operations, or repeated applications of triple-operations, are regarded as operators which induce transformations of distance matrices and the operator semigroup is defined. In §3, an algorithm, as well as an operator, is regarded as a systematic enumeration of paths in a network and the correspondence between algorithms and path-sets is established. Based on that correspondence the validity and optimality of an algorithm are discussed. It is shown that any algorithm valid for nonnegative distance matrices is valid for general distance matrices, and that any valid algorithm for an  $n$ -node complete network should have at least  $n(n-1)(n-2)$  occurrences

of triple-operations. This last proposition leads to the proof of the optimality of many existing algorithms such as the Warshall-Floyd algorithm, the Dantzig algorithm, the Katayama-Watanabe algorithm, etc. Also the algebraic structures of a class of algorithms are investigated from the points of view of equivalence relation, partial order, etc. Useful formulae for the formal manipulation of algorithms are listed, by means of which the validity of various algorithms is proved. In §4, some specially structured networks, such as cascade networks, star networks, etc., are treated, where some of the existing algorithms for such networks are shown to be nonoptimal and new optimal algorithms are proposed.

Most of the results described in the present paper are based on our preliminary studies reported separately in [1], [2] and [3].

## 2. *The All-Shortest-Path Algorithms*

Let us consider a network  $G$  with  $n$  nodes. We denote the set of nodes by  $N = \{1, 2, \dots, n\}$ . We assume that  $G$  has no parallel branches.

We define a *path* from node  $i_0$  to node  $i_r$  of length  $r$  in  $G$  as a sequence  $(i_0, i_1, \dots, i_{r-1}, i_r)$  of  $r + 1$  nodes of  $G$ .

If  $i_0 = i_r$ , we say that the path is *closed*.

For each branch connecting two nodes, say  $i$  and  $j$ , we denote by  $d_{ij}$  the *span* (or *distance*) of the branch measured from  $i$  to  $j$ . [It may be  $d_{ij} \neq d_{ji}$  or  $d_{ij} < 0$  for some pairs  $(i, j)$ .] If two distinct nodes  $i$  and  $j$  are connected by no branch, we set  $d_{ij} = d_{ji} = \infty$ . For every node  $i$  we set  $d_{ii} = 0$ . We call the sum of  $d_{ij}$ 's along a path the *span* (or *distance*) of the path (we assume the span of a path of length 0 to be 0).

In the following the *span of any closed path is assumed to be nonnegative*, since, otherwise, the problem would have no solution. We call a matrix  $D = (d_{ij})$  of spans  $d_{ij}$ 's satisfying this condition a *distance matrix on G*.

Let  $D = (d_{ij})$  be a distance matrix on  $G$ . We define the *shortest distance*  $\bar{d}_{ij}^\infty$  from node  $i$  to node  $j$  corresponding to  $D$  as the minimum of the spans of all the paths (of any length) from  $i$  to  $j$ , and call the path(s) giving  $\bar{d}_{ij}^\infty$  the *shortest path(s)* from  $i$  to  $j$  corresponding to  $D$ . We call the matrix  $D^\infty = (\bar{d}_{ij}^\infty)$  the *shortest-distance matrix corresponding to D*. The *all-shortest-path problem* on a network  $G$  is the problem of obtaining the shortest distance matrix  $D^\infty = (\bar{d}_{ij}^\infty)$  corresponding to an arbitrary given distance matrix  $D = (d_{ij})$  on  $G$ .

Numerous methods of solution have been proposed for the all-shortest-path problem. Most of them are expressed as repeated applications of the so-called triple-operations, where

a triple-operation with pivot  $k_0$  on pair  $(i_0, j_0)$  (to be denoted by  $\langle i_0 j_0 \rangle^{k_0}$  in abbreviation) of an  $n$  by  $n$  matrix  $A = (a_{ij})$  transforms  $A$  into  $B = (b_{ij})$  defined as

$$\left. \begin{aligned} b_{i_0 j_0} &= \min (a_{i_0 j_0}, a_{i_0 k_0} + a_{k_0 j_0}) ; \\ b_{ij} &= a_{ij} \quad \text{where } (i, j) \neq (i_0, j_0) , \end{aligned} \right\} \quad (2.1)$$

or, in ALGOL-like expression,  $\langle i_0 j_0 \rangle^{k_0}$  has the effect:

$$a_{i_0 j_0} := \min(a_{i_0 j_0}, a_{i_0 k_0} + a_{k_0 j_0}) \quad (2.2)$$

for the array  $A = (a_{ij})$ . Using this kind of expression, some of the well-known methods of solution for the all-shortest-path problem may be described as follows.

*Warshall-Floyd method* [4]:

```

for k := 1 step 1 until n do
  for i := 1 step 1 until n do
    for j := 1 step 1 until n do
      aij := min (aij, aik + akj)

```

*Dantzig method* [5]:

```

for k := 1 step 1 until n do
  begin for i := 1 step 1 until k do
        for l := 1 step 1 until k do
          aik := min(aik, ail + alk);
        for j := 1 step 1 until k do
          for l := 1 step 1 until k do

```

```

     $a_{kj} := \min (a_{ij}, a_{kl} + a_{lj});$ 
    for  $i := 1$  step 1 until  $k$  do
    for  $j := 1$  step 1 until  $k$  do
     $a_{ij} := \min (a_{ij}, a_{ik} + a_{kj})$ 
end

```

*Katayama-Watanabe method* [6]:

```

for  $i := 1$  step 1 until  $n$  do
for  $j := 1$  step 1 until  $n$  do
for  $k := 1$  step 1 until  $\min(i, j)$  do
 $a_{ij} := \min (a_{ij}, a_{ik} + a_{kj});$ 
for  $i := n$  step -1 until 1 do
for  $j := n$  step -1 until 1 do
for  $k := \max(i, j)$  step 1 until  $n$  do
 $a_{ij} := \min (a_{ij}, a_{ik} + a_{kj});$ 
for  $i := 1$  step 1 until  $n$  do
for  $j := 1$  step 1 until  $n$  do
for  $k := \min(i, j)$  step 1 until  $\max(i, j)$  do
 $a_{ij} := \min (a_{ij}, a_{ik} + a_{kj})$ 

```

*Cascade method* [7], [8]:

```

for  $i := 1$  step 1 until  $n$  do
for  $j := 1$  step 1 until  $n$  do
for  $k := 1$  step 1 until  $n$  do
 $a_{ij} := \min (a_{ij}, a_{ik} + a_{kj});$ 

```

```

for i := n step -1 until 1 do
for j := n step -1 until 1 do
for k := 1 step 1 until n do
aij := min (aij, aik + akj)

```

Doubling power method with successive replacement (see, e.g., [9]):

```

for l := 1 step 1 until 3 do
for i := 1 step 1 until n do
for j := 1 step 1 until n do
for k := 1 step 1 until n do
aij := min (aij, aik + akj)

```

In the following we shall mean by an *algorithm* any sequence of triple-operations. Thus, an algorithm is expressed as

$$\langle i_t^k j_t \rangle \cdot \langle i_{t-1}^{k_{t-1}} j_{t-1} \rangle \cdots \langle i_2^{k_2} j_2 \rangle \cdot \langle i_1^{k_1} j_1 \rangle \quad (t \geq 0), \quad (2.3)$$

where the triple-operations are performed successively from right to left. For two algorithms  $\tau$  and  $\tau'$  we denote by  $\tau \cdot \tau'$  the algorithm in which the algorithm  $\tau'$  is performed first and then,  $\tau$ . For an algorithm  $\tau$  and a distance matrix  $D$ , we denote by  $\tau D$  the result of  $\tau$  operating on  $D$ .

We say that an algorithm  $\tau$  is *valid* for a network  $G$  if  $\tau D = D^\infty$  for every distance matrix on  $G$  (which means that we allow "invalid" algorithms also as algorithms).

We define the *complexity* of an algorithm as the number of occurrences of triple-operations in the algorithm, i.e. the value of " $t$ " in (2.3). We denote the algorithm of complexity 0 by  $\varepsilon$ , which induces the identity transformation on distance matrices.

We say that a valid algorithm for  $G$  is *optimal* if its complexity is the smallest of all the valid algorithms for  $G$ .

### 3. The Operator Semigroup and Path-Sets

3.1. *Enumeration of paths by an algorithm.* We begin with the investigation into algorithms from the point of view of enumeration of paths. We call a set of paths a *path-set*. We call the subset of a path-set  $\pi$ , which consists of the paths from node  $i$  to node  $j$  in  $\pi$ , the  $(i, j)$ -*component* of  $\pi$  and denote it by  $\pi_{ij}$ . Sometimes we denote a path-set  $\pi$  in a matrix form as  $\pi = (\pi_{ij})$ . We denote by  $P = (P_{ij})$  the path-set of all the paths in  $G$  and by  $P^{(r)} = (P_{ij}^{(r)})$  the path-set of all the paths of length not exceeding  $r$  in  $G$ .

For two paths  $p = (i_0, i_1, \dots, i_{r-1}, i_r)$  and  $q = (j_0, j_1, \dots, j_{s-1}, j_s)$ , we define the *concatenation*  $p \square q$  as the path  $(i_0, i_1, \dots, i_{r-1}, i_r, j_1, \dots, j_{s-1}, j_s)$  if  $i_r = j_0$  (if  $i_r \neq j_0$ ,  $p \square q$  is undefined). For two path-sets  $\pi$  and  $\pi'$ , we define the concatenation as  $\pi \square \pi' = \{p \square p' \mid p \in \pi, p' \in \pi'\}$ . For a path



$p = (i_0, i_1, \dots, i_{r-1}, i_r)$  and a path-set  $\pi = (\pi_{ij})$ , we define the path-set  $p[\pi]$  by  $p[\pi] = \pi_{i_0 i_1} \square \pi_{i_1 i_2} \square \dots \square \pi_{i_{r-1} i_r}$  if  $r \geq 1$ , or by  $p[\pi] = \{p\}$  if  $r = 0$ . For path-sets  $\pi$  and  $\pi'$ , we define the *link-product*  $\pi \uparrow \pi'$  by  $\pi \uparrow \pi' = \bigcup_{p \in \pi} p[\pi']$ .

To each algorithm  $\tau$  we assign a path-set  $\rho(\tau)$  as follows:

$$\left. \begin{aligned} \rho(\epsilon) &= P^{(1)} ; \\ \rho(\langle i^k j \rangle) &= P^{(1)} \cup \{(i, k, j)\} ; \\ \rho(\tau \cdot \tau') &= \rho(\tau) \uparrow \rho(\tau') . \end{aligned} \right\} \quad (3.1)$$

The intuitive interpretation of  $\rho$  is given by the following theorem, which is easy to prove.

*Theorem 1. For a distance matrix  $D$  and an algorithm  $\tau$ , each  $(i, j)$ -entry of the resultant matrix  $\tau D$  is equal to the minimum of the spans along all the paths from node  $i$  to node  $j$  in the path-set  $\rho(\tau)$ .*

3.2. *Validity of an algorithm.* A path is said to be *elementary* if the nodes on the path are distinct. We denote by  $\rho^E(\tau)$  the path-set of all elementary paths in the path-set  $\rho(\tau)$ . We say that a path is *conductive* if each pair of neighboring nodes in the path is connected by a branch.

It is obvious that the shortest path from a node to another corresponding to any distance matrix is elementary and conductive.

It is also obvious that any elementary and conductive path from a node to another may be the shortest path corresponding to some distance matrix. Hence follows the

*Theorem 2. An algorithm  $\tau$  is valid for a network  $G$  if and only if all the elementary conductive paths in  $G$  are contained in  $\rho^\varepsilon(\tau)$ .*

3.3. *Optimality of an algorithm.* A network is said to be *complete* if every pair of different nodes is connected by a branch.

In a complete network  $G$ , all the elementary paths (and, therefore, all the elementary paths of length 2) are conductive. If a triple-operation  $\langle i^k j \rangle$  ( $i, j, k$  being distinct) did not occur in an algorithm  $\tau$ , then the elementary conductive path  $(i, k, j)$  would not be contained in  $\rho^\varepsilon(\tau)$ . Hence we have

*Theorem 3. For every triple  $(i, j, k)$  of distinct nodes  $i, j$  and  $k$ , the triple-operation  $\langle i^k j \rangle$  occurs at least once in any valid algorithm for a complete network.*

and

*Theorem 4. Any valid algorithm for a complete  $n$ -node network contains at least  $n(n-1)(n-2)$  occurrences of triple-operations.*

Theorem 4 shows that the methods by R. W. Floyd [4], G. B.

Dantzig [5], and H. Katayama and H. Watanabe [6], are optimal for a complete network in this sense.

It is important to note that the above-mentioned fact is independent of whether the distance matrix under consideration is nonnegative or not, i.e.

*Theorem 5. If an algorithm transforms every nonnegative distance matrix on a network (not necessarily complete) into the corresponding shortest distance matrix, then the algorithm is valid for that network, i.e. it transforms every distance matrix (not necessarily nonnegative) on the network into the corresponding shortest distance matrix.*

and

*Theorem 6. On a complete  $n$ -node network no algorithm which consist of less than  $n(n-1)(n-2)$  triple-operations (therefore, invalid for a complete  $n$ -node network) transforms every nonnegative distance matrix into the corresponding shortest distance matrix.*

Recently, A. J. Hoffman and S. Winograd [10] devised a method by which only  $O(n^{5/2})$  additions and subtractions are needed. However, as for the operations of comparison the method is identical with that of Floyd, i.e. it requires about  $n^3$  comparisons. Moreover, the method is not an algorithm in our sense, since it makes use of the combinations of addition/subtraction and comparison in a form different from triple-operations.

3.4. *The operator semigroup.* Let  $\pi$  and  $\pi'$  be path-sets of elementary paths in a network  $G$ . If all the conductive paths in  $\pi$  are contained also in  $\pi'$ , we write  $\pi \subseteq \pi'$  (C). Let  $\tau$  and  $\tau'$  be algorithms. We write  $\tau \leq \tau'$  if  $\rho^\varepsilon(\tau) \supseteq \rho^\varepsilon(\tau')$  (C). We say that  $\tau$  and  $\tau'$  are *equivalent* and write  $\tau \equiv \tau'$  if  $\tau \leq \tau'$  and  $\tau' \leq \tau$ . We write  $\tau < \tau'$  if  $\tau \not\equiv \tau'$  and  $\tau \leq \tau'$ . We say that  $\tau$  and  $\tau'$  are *incomparable* and write  $\tau \not\sim \tau'$  if neither  $\tau \leq \tau'$  nor  $\tau' \leq \tau$ .

The quotient of the set of algorithms by the equivalence  $\equiv$  defined above forms a semigroup, which we shall call the *operator semigroup*. Each member of the operator semigroup (i.e. each equivalence class of algorithms) is called an *operator*.

Let us examine some properties of the operator semigroup.

To begin with, we observe that

$$\langle i \ i \rangle^k \equiv \langle k \ j \rangle^k \equiv \langle i \ k \rangle^k \equiv \varepsilon \quad \text{for every } i, j, k \text{ of } N, \quad (3.2)$$

and

$$\langle i \ j \rangle^k \cdot \langle i \ j \rangle^k \equiv \langle i \ j \rangle^k \quad \text{for every } i, j, k \text{ of } N. \quad (3.3)$$

The commutativity relations between triple-operations are summarized in Table 1, where  $\langle i \ j \rangle^k = \tau$  and  $\langle i' \ j' \rangle^{k'} = \tau'$ . The inequalities  $\leq$  in Table 1 hold strictly (i.e. as  $<$ ) if the underlying network  $G$  is complete.

	$j = j'$		$j \neq j'$		
$i = i'$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$		$k' = j$	$k' \neq j$	
			$k = j'$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \leq \tau' \cdot \tau$
			$k \neq j'$	$\tau \cdot \tau' \geq \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$
$i \neq i'$		$k' = i$	$k' \neq i$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	
	$k = i'$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \leq \tau' \cdot \tau$		
	$k \neq i'$	$\tau \cdot \tau' \geq \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$		

Table 1. Commutativity relations between triple-

operations  $\tau = \langle i^k_j \rangle$  and  $\tau' = \langle i'j' \rangle$ .

3.5. *Operators of practical use.* From the fundamental properties of the operator semigroup shown in §3.4, we see that

$$\langle S^k \rangle \stackrel{d}{=} \prod_{(i,j) \in S} \langle i^k_j \rangle \quad (k \in N ; S \subseteq N \times N)$$

has its unique meaning as an operator, where the right-hand side means the "."-product of the triple-operations of the form  $\langle i^k_j \rangle$   $[(i, j) \in S]$ , the order being unimportant. It is easy to prove that

$$\langle R^k \rangle \cdot \langle S^k \rangle \equiv \langle R \cup S^k \rangle \quad \text{for every } k \in N \text{ and every } R, S \subseteq N \times N. \tag{3.4}$$

The following symbols for subsets of  $N \times N$  will be used.

$$\begin{aligned}
 I(k) &= \{(i, j) \in N \times N \mid i \leq k \leq j\}, \\
 II(k) &= \{(i, j) \in N \times N \mid i \leq k, j \leq k\}, \\
 III(k) &= \{(i, j) \in N \times N \mid j \leq k \leq i\}, \\
 IV(k) &= \{(i, j) \in N \times N \mid k \leq i, k \leq j\}.
 \end{aligned}
 \tag{3.5}$$

By an elementary consideration we can prove the commutativity relations on a complete network in Table 2 between operators of the form  $\langle R(k) \rangle$  and  $\langle S(k') \rangle$  ( $R, S = I, II, III$  or  $IV$ ), where the commutativity relation between  $\langle I(k) \rangle$ 's or  $\langle III(k) \rangle$ 's, respectively, follows from the fact that

$$\begin{aligned}
 \langle i \ j \rangle \cdot \langle k \ j \rangle \cdot \langle i \ k' \rangle \cdot \langle i \ k \rangle &\equiv \langle i \ j \rangle \cdot \langle i \ k \rangle \cdot \langle i \ j \rangle \cdot \langle k' \ j \rangle \\
 &\text{for every } i, j, k \text{ and } k' \in N,
 \end{aligned}
 \tag{3.6}$$

which can be examined directly.

We define

	$S = I$	$S = II$	$S = III$	$S = IV$
$R = I$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \not\equiv \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$
$R = II$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' < \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$
$R = III$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \not\equiv \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$	$\tau \cdot \tau' \equiv \tau' \cdot \tau$
$R = IV$	$\tau \cdot \tau' \not\equiv \tau' \cdot \tau$	$\tau \cdot \tau' \not\equiv \tau' \cdot \tau$	$\tau \cdot \tau' \not\equiv \tau' \cdot \tau$	$\tau \cdot \tau' > \tau' \cdot \tau$

Table 2. Commutativity relations on a complete network between operators  $\tau = \langle R(k) \rangle$  and  $\tau' = \langle S(k') \rangle$ , where  $k < k'$ .

$$\left. \begin{aligned} \Gamma_r^F &\doteq \langle R(n) \rangle \cdot \langle R(n-1) \rangle \cdots \langle R(2) \rangle \cdot \langle R(1) \rangle, \\ \Gamma_r^B &\doteq \langle R(1) \rangle \cdot \langle R(2) \rangle \cdots \langle R(n-1) \rangle \cdot \langle R(n) \rangle, \end{aligned} \right\} \quad (3.7)$$

where we set  $r = 1, 2, 3$  or  $4$ , according as  $R = I, II, III$  or  $IV$ , respectively. If  $\langle R(k) \rangle$ 's are commutative for different  $k$ 's, we write simply as  $\Gamma_r$  (cf. Table 2). We sometimes use such notations as  $\Gamma_{rs}^F$ ,  $\Gamma_{rs}^B$  or  $\Gamma_{rst}^F$ ,  $\Gamma_{rst}^B$  to mean a product of operators  $\langle R(k) \mathbf{U}_S(k) \rangle$ 's or that of  $\langle R(k) \mathbf{U}_S(k) \mathbf{U}_T(k) \rangle$ 's, whose rigorous definition will be evident. Thus the following notations have their unique meanings as operators.

$$\begin{array}{cccccccc} \Gamma_1, & \Gamma_3, & \Gamma_{13}, & & & & & \\ \Gamma_2^B, & \Gamma_2^F, & \Gamma_{12}^B, & \Gamma_{12}^F, & \Gamma_{23}^B, & \Gamma_{23}^F, & \Gamma_{123}^B, & \Gamma_{123}^F, \\ \Gamma_4^B, & \Gamma_4^F, & \Gamma_{34}^B, & \Gamma_{34}^F, & \Gamma_{14}^B, & \Gamma_{14}^F, & \Gamma_{134}^B, & \Gamma_{134}^F. \end{array}$$

The intuitive interpretation of these operators are given in Fig. 1, where the shaded area indicates the set of the pairs of lower indices of triple-operations with the pivot indicated by a small circle and the arrow indicates the movement of the pivot.

From the properties in Table 2 we can obtain various relations among the above operators. For example, we have

$$\left. \begin{aligned} \Gamma_{13} &\equiv \Gamma_1 \cdot \Gamma_3 \equiv \Gamma_3 \cdot \Gamma_1, \\ \Gamma_{12}^B &\equiv \Gamma_1 \cdot \Gamma_2^B, & \Gamma_{12}^F &\equiv \Gamma_2^F \cdot \Gamma_1, \end{aligned} \right\} \quad (3.8)$$

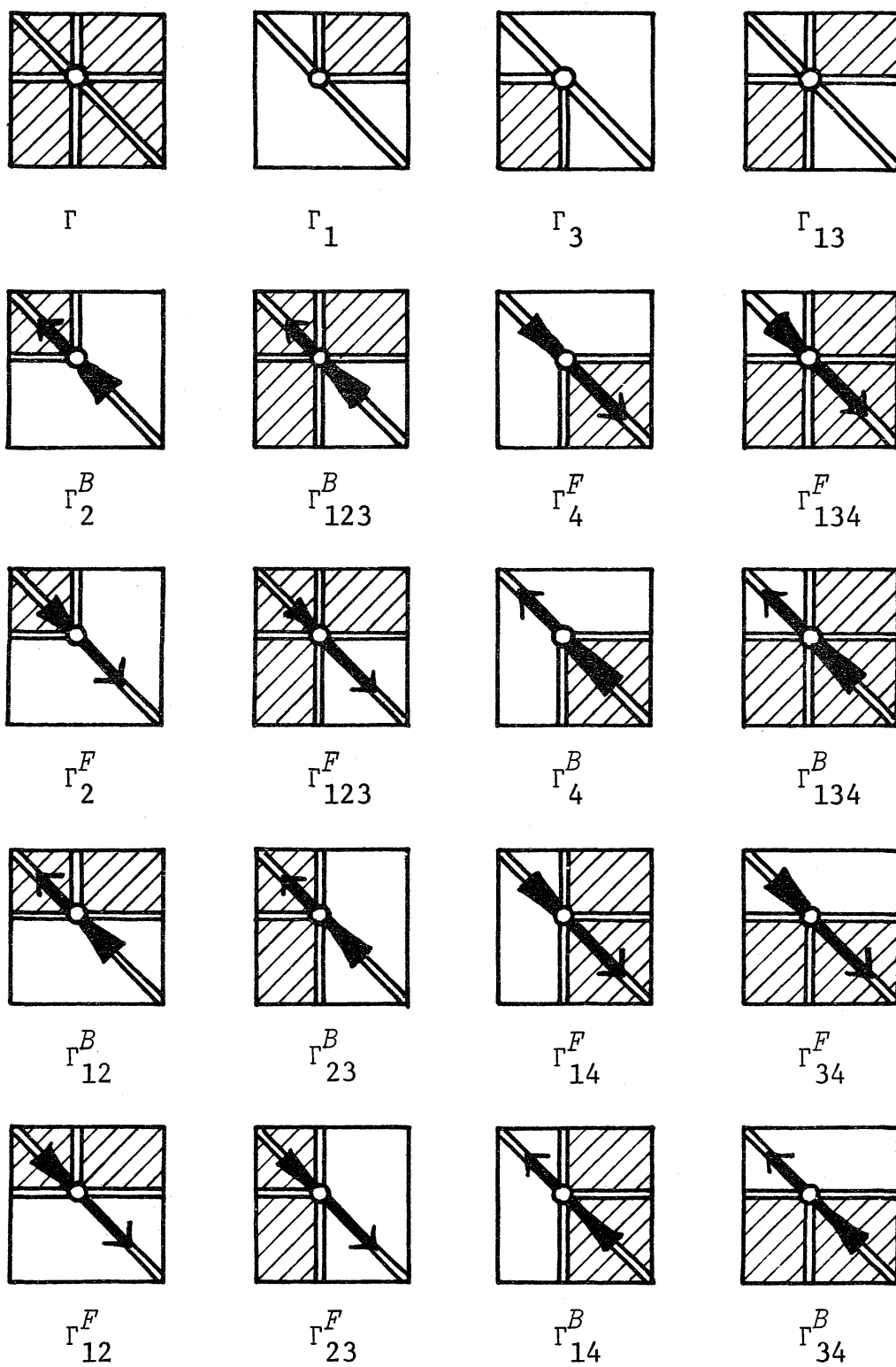


Fig. 1



and

$$\Gamma_2^B < \Gamma_2^F \quad \text{if the underlying network is complete.} \quad (3.9)$$

(For further details, see [2].)

Some of these operators are characterized in terms of the corresponding path-sets as follows.

$\rho^\varepsilon(\Gamma_1)$  consists of all the elementary paths of lengths 0 and 1 and of the paths  $(i_0, i_1, \dots, i_{r-1}, i_r)$ 's such that  $i_0 < i_1 < \dots < i_{r-1} < i_r$ .

$\rho^\varepsilon(\Gamma_2^B)$  consists of all the elementary paths of lengths 0 and 1 and of the paths  $(i_0, i_1, \dots, i_{r-1}, i_r)$ 's such that  $i_t > i_0$  and  $i_t > i_r$  for  $t = 1, \dots, r-1$ .

$\rho^\varepsilon(\Gamma_3)$  consists of all the elementary paths of lengths 0 and 1 and of the paths  $(i_0, i_1, \dots, i_{r-1}, i_r)$ 's such that  $i_0 > i_1 > \dots > i_{r-1} > i_r$ .

$\rho^\varepsilon(\Gamma_4^F)$  consists of all the elementary paths of lengths 0 and 1 and of the paths  $(i_0, i_1, \dots, i_{r-1}, i_r)$ 's such that  $i_t < i_0$  and  $i_t < i_r$  for  $t = 1, \dots, r-1$ .

3.6. *Variants of Warshall-Floyd algorithm.* The path-set

$\rho^\varepsilon(\langle \begin{smallmatrix} k \\ N \times N \end{smallmatrix} \rangle)$  consists of all the elementary paths of lengths 0 and 1 and of the paths of the form  $(i, k, j)$  ( $i, j = 1, 2, \dots, n$ ).

The path-set  $\rho^\varepsilon(\langle N \times N \rangle^{k_1} \dots \langle N \times N \rangle^{k_s})$  includes all the elementary paths  $(i, i_1, \dots, i_{r-1}, j)$ 's ( $r = 1, 2, \dots, s + 1; i_1, \dots, i_{r-1} = k_1, \dots, k_s; i, j = 1, 2, \dots, n$ ). Therefore,

$$\Gamma \stackrel{d}{=} \prod_{k \in N} \langle N \times N \rangle^k \quad (3.10)$$

has its unique meaning as an operator valid for a complete network. This operator  $\Gamma$  represents the Warshall-Floyd algorithm [4] which has been referred to in §2.

From §3.6 we have various equivalent expressions for  $\Gamma$  such as follows.

$$\begin{aligned} \Gamma &\equiv \Gamma_{123}^F \cdot \Gamma_4^F \equiv \Gamma_{123}^B \cdot \Gamma_4^F \equiv \Gamma_{134}^B \cdot \Gamma_2^B \equiv \Gamma_{134}^F \cdot \Gamma_2^B \\ &\equiv \Gamma_{12}^F \cdot \Gamma_{34}^F \equiv \Gamma_{12}^B \cdot \Gamma_{34}^F \equiv \Gamma_{14}^B \cdot \Gamma_{23}^B \equiv \Gamma_{14}^F \cdot \Gamma_{23}^B \\ &\equiv \Gamma_{23}^F \cdot \Gamma_{14}^F \equiv \Gamma_{23}^B \cdot \Gamma_{14}^F \equiv \Gamma_{34}^B \cdot \Gamma_{12}^B \equiv \Gamma_{34}^F \cdot \Gamma_{12}^B \\ &\equiv \Gamma_2^F \cdot \Gamma_{13} \cdot \Gamma_4^F \equiv \Gamma_2^B \cdot \Gamma_{13} \cdot \Gamma_4^F \equiv \Gamma_4^B \cdot \Gamma_{13} \cdot \Gamma_2^B \equiv \Gamma_4^F \cdot \Gamma_{13} \cdot \Gamma_2^B \\ &\equiv \Gamma_{13} \cdot \Gamma_2^B \cdot \Gamma_4^F \equiv \Gamma_{13} \cdot \Gamma_4^F \cdot \Gamma_2^B \\ &\equiv \Gamma_1 \cdot \Gamma_2^B \cdot \Gamma_3 \cdot \Gamma_4^F \equiv \Gamma_3 \cdot \Gamma_2^B \cdot \Gamma_1 \cdot \Gamma_4^F \\ &\equiv \Gamma_1 \cdot \Gamma_4^F \cdot \Gamma_3 \cdot \Gamma_2^B \equiv \Gamma_3 \cdot \Gamma_4^F \cdot \Gamma_1 \cdot \Gamma_2^B \end{aligned} \quad (3.11)$$

These expressions for  $\Gamma$  will be useful to devise less complex algorithms in the case where the underlying network is of special structure (cf. §4).

3.7. Another proof of the validity of Katayama-Watanabe algorithm. Table 1 shows that

$$\langle i \ j \rangle^X \stackrel{d}{=} \prod_{k \in X} \langle i \ j \rangle^k$$

has its unique meaning as an operator. It is sometimes convenient to adopt the following subsets of  $N$  as the  $X$  in the above operator  $\langle i \ j \rangle^X$ .

$$\left. \begin{aligned} I(i, j) &= \{k \in N \mid i \leq k \leq j\}, \\ II(i, j) &= \{k \in N \mid i \leq k, j \leq k\}, \\ III(i, j) &= \{k \in N \mid j \leq k \leq i\}, \\ IV(i, j) &= \{k \in N \mid k \leq i, k \leq j\}. \end{aligned} \right\} \quad (3.12)$$

In the same way as we defined  $\Gamma$ -operators in §3.6, we define

$$\left. \begin{aligned} \gamma_r^F &\stackrel{d}{=} \gamma_r^{F(n)} \cdots \gamma_r^{F(1)}, & \gamma_r^{F(i)} &\stackrel{d}{=} \langle R(i, n) \rangle \cdots \langle R(i, 1) \rangle, \\ \gamma_r^B &\stackrel{d}{=} \gamma_r^{B(1)} \cdots \gamma_r^{B(n)}, & \gamma_r^{B(i)} &\stackrel{d}{=} \langle R(i, 1) \rangle \cdots \langle R(i, n) \rangle, \end{aligned} \right\} \quad (3.13)$$

where we set  $r = 1, 2, 3$  or  $4$ , according as  $R = I, II, III$  or  $IV$ , respectively. We sometimes use symbols such as  $\gamma_{rs}^F, \gamma_{rs}^B$  or  $\gamma_{rst}^F, \gamma_{rst}^B$  to mean the product operators of  $\langle R(i, j) \mathbf{US}(i, j) \rangle_s$  or  $\langle R(i, j) \mathbf{US}(i, j) \mathbf{UT}(i, j) \rangle_j$ 's, whose rigorous definition will be

evident. For example, we may make use of

$$\begin{aligned} &\gamma_2^B, \quad \gamma_2^F, \quad \gamma_4^B, \quad \gamma_4^F, \\ &\gamma_1^B, \quad \gamma_1^F, \quad \gamma_3^B, \quad \gamma_3^F, \quad \gamma_{13}^B, \quad \gamma_{13}^F, \end{aligned}$$

$$\begin{array}{cccccccc} \gamma_2^B & \gamma_2^F & \gamma_{12}^B & \gamma_{12}^F & \gamma_{23}^B & \gamma_{23}^F & \gamma_{123}^B & \gamma_{123}^F \\ \gamma_4^B & \gamma_4^F & \gamma_{34}^B & \gamma_{34}^F & \gamma_{14}^B & \gamma_{14}^F & \gamma_{134}^B & \gamma_{134}^F \end{array}$$

By an elementary computation we can derive various formulae connecting  $\Gamma$ -operators and  $\gamma$ -operators. For example, we have (see [2])

$$\left. \begin{array}{ll} \gamma_1^B \equiv \gamma_1^F \equiv \Gamma_1, & \gamma_{13}^B \equiv \gamma_{13}^F \equiv \Gamma_{13}; \\ \gamma_2^F \equiv \Gamma_2^F, & \gamma_2^B \equiv \Gamma_2^B; \\ \gamma_{12}^B \equiv \Gamma_2^B \cdot \Gamma_1, & \end{array} \right\} \quad (3.14)$$

and

$$\gamma_{12}^F > \Gamma_1 \cdot \Gamma_2^F, \quad \gamma_{12}^F \neq \Gamma_{12}^F, \quad \gamma_{12}^B > \Gamma_{12}^B$$

*if the underlying network is complete.* (3.15)

The validity of Katayama-Watanabe algorithm can be proved on the basis of the preceding relations among operators as follows. Their algorithm is described in terms of  $\gamma$ -operators as  $\gamma_{13}^F \cdot \gamma_2^B \cdot \gamma_4^F$ , and, by virtue of (3.14), this is equivalent to  $\Gamma_{13}^F \cdot \Gamma_2^B \cdot \Gamma_4^F$ . However, we have already obtained in §3.6 this last expression as a variant of Warshall-Floyd algorithm  $\Gamma$ .

The validity of other methods mentioned in §2 may be proved in a similar way (see [2]), and the process of proof is far simpler than the direct proof given in the original papers.

#### 4. Optimal Algorithms for Specially Structured Networks

##### 4.1. A general lower bound for the complexity of algorithms.

In order to discuss the optimality of algorithms for specially structured networks, we establish the following preliminary theorems.

*Theorem 7.* If a path  $p = (i, k_1, \dots, k_{r-1}, j)$  ( $r \geq 2$ ) belongs to the path-set  $\rho^\varepsilon(\tau)$  for an algorithm  $\tau$ , then there exists at least one node  $k_t$  ( $1 \leq t \leq r - 1$ ) such that  $\langle i \begin{smallmatrix} k_t \\ j \end{smallmatrix} \rangle$  appears at least once in  $\tau$ .

*Proof:* Suppose none of  $\langle i \begin{smallmatrix} k_1 \\ j \end{smallmatrix} \rangle, \dots, \langle i \begin{smallmatrix} k_{r-1} \\ j \end{smallmatrix} \rangle$  appear in  $\tau$ . Then  $p$  would not belong to  $\rho^\varepsilon(\tau)$ , contrary to the assumption. ■

We say that two elementary paths  $p = (i, k_1, \dots, k_{r-1}, j)$  ( $r \geq 2$ ) and  $q = (i, h_1, \dots, h_{s-1}, j)$  ( $s \geq 2$ ) are *disjoint* if nodes  $k_1, \dots, k_{r-1}$  are not on  $q$  and  $h_1, \dots, h_{s-1}$  are not on  $p$ . For two distinct nodes  $i$  and  $j$  we denote by  $u(i, j)$  the maximum number of disjoint elementary conductive paths of length greater than 1 from  $i$  to  $j$ . From Theorem 7 we have

*Theorem 8.* Let  $i$  and  $j$  be arbitrary distinct nodes and  $p_1, p_2, \dots, p_{u(i, j)}$  be disjoint elementary conductive paths from  $i$  to  $j$  in a network  $G$ . If  $\tau$  is a valid algorithm for  $G$ , then for each  $p_v$  [ $v = 1, 2, \dots, u(i, j)$ ] there is at least

one  $k_v$  (distinct from  $i$  and  $j$ ) on  $p_v$  such that  $\langle i \ j \ k_v \rangle$  appears at least once in  $\tau$ .

and

*Theorem 9.* Any valid algorithm for a network  $G$  contains at least  $\sum_{i \neq j} u(i, j)$  occurrences of triple-operations, where the summation is taken over all the pairs of nodes.

*Remark 1.* Note that, by virtue of a well-known theorem of the Menger type [12], the number  $u(i, j)$  is equal to the minimum number of nodes whose removal separates nodes  $i$  and  $j$  from each other.

*Remark 2.* If  $G$  is complete, then  $u(i, j) = n - 2$  for every pair  $(i, j)$ , so that we have the lower bound  $n(n-1)(n-2)$ , in accordance with the arguments in the preceding section.

4.2. *Tree-structured networks.* For a partition of the set of nodes  $N$  of a network  $G$  into  $m$  disjoint subsets  $N_1, N_2, \dots, N_m$ , we construct the graph  $\tilde{G}$  which has  $N_1, N_2, \dots, N_m$  as the nodes and which has a branch connecting  $N_v$  and  $N_w$  ( $v \neq w$ ) if and only if there is a branch in  $G$  connecting a node of  $N_v$  and a node of  $N_w$ . If  $\tilde{G}$  is a tree and if there is always a branch in  $G$  connecting nodes  $i$  and  $j$  when  $i, j \in N_v$  or  $i \in N_v, j \in N_w$  with  $N_v$  and  $N_w$  adjacent, we say that the network  $G$  is *tree-structured*. An

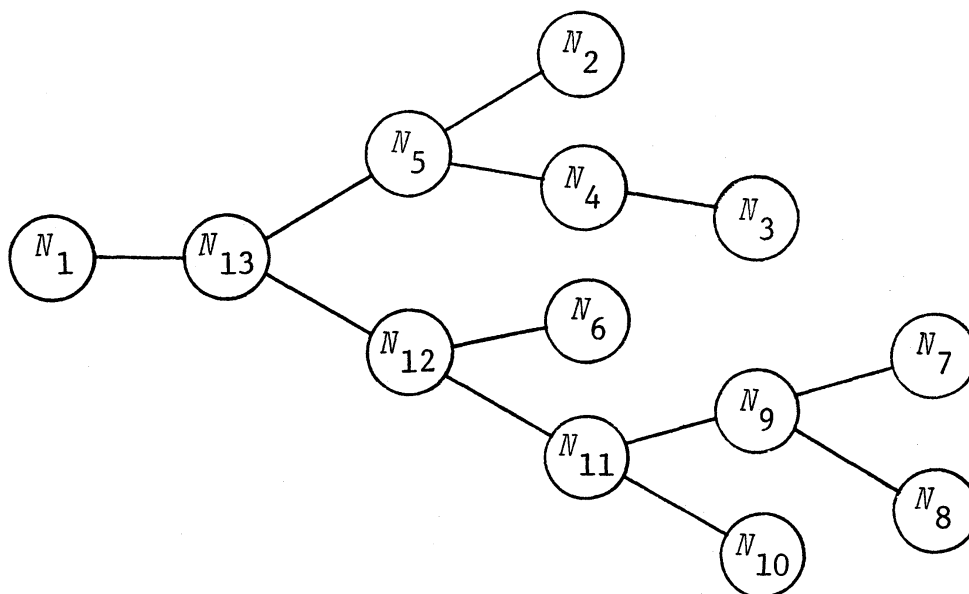


Fig. 2.  $\tilde{G}_{tr}$  of a tree-structured network  $G_{tr}$ .

example of a tree-structured network is shown in Fig. 2.

Let  $G_{tr}$  be a tree-structured network. We say that two node sets  $N_v$  and  $N_w$  are *adjacent* in  $G_{tr}$  if nodes  $N_v$  and  $N_w$  in  $\tilde{G}_{tr}$  are adjacent in  $\tilde{G}_{tr}$ . We say that a node set  $N_x$  is *intermediate* from  $N_v$  to  $N_w$  if every conductive path from a node in  $N_v$  to a node in  $N_w$  passes a node in  $N_x$ .

Theorem 8, when applied to tree-structured networks, gives

*Theorem 10. Let  $\tau$  be a valid algorithm for a tree-structured network.*

- (a) *If two distinct nodes  $i$  and  $j$  belong to the same node set  $N_v$ , then for every node  $k$  in  $N_v$  and all the node sets adjacent to  $N_v$ , the triple-operation  $\langle i^k_j \rangle$  appears at least once in  $\tau$ .*

- (b) If two node sets  $N_v$  and  $N_w$  are adjacent and  $i \in N_v$ ,  $j \in N_w$ , then for every node  $k$  in  $N_v$  and  $N_w$  the triple-operation  $\langle i^k j \rangle$  appears at least once in  $\tau$ .
- (c) If two node sets  $N_v$  and  $N_w$  are not adjacent and  $i \in N_v$ ,  $j \in N_w$ , then there exists an intermediate node set  $N_x$  such that for every node  $k$  in  $N_x$  the triple-operation  $\langle i^k j \rangle$  appears at least once in  $\tau$ .

Quite recently, D. R. Shier presented a method of solution for a tree-structured networks [13]. When we express this method in terms of triple-operations, we obtain an algorithm containing only those triple-operations which are mentioned in Theorem 10, each triple-operation occurring precisely once. Moreover, for nodes  $i$  and  $j$  ( $i \in N_v$ ,  $j \in N_w$ ,  $N_v$  and  $N_w$  not being adjacent) we can choose  $N_x$  in (c) of Theorem 10 such that  $|N_x|$  is the minimum among all the  $N_x$ 's intermediate from  $N_v$  to  $N_w$  (see §4.3). Thus, when appropriately programmed, his algorithm is optimal.

4.3. *Cascade networks.* As a special case of tree-structured networks, a network such as in Fig. 3 may be considered. However,

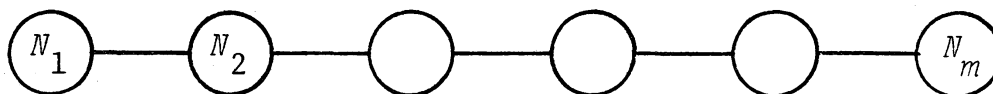


Fig. 3



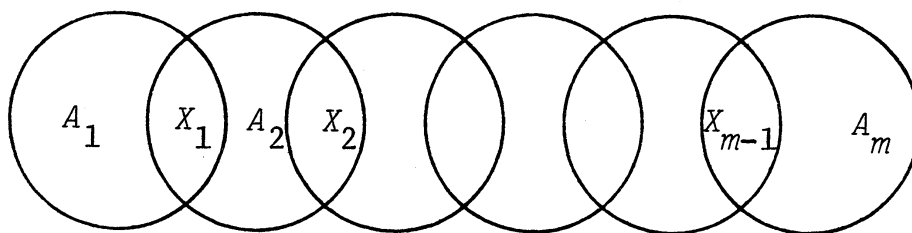


Fig. 4. A cascade network.

in the following we shall consider "cascade networks" such as illustrated in Fig. 4 by slightly generalizing it.

Let the set of nodes  $N$  be decomposed into disjoint subsets  $A_1, X_1, A_2, X_2, \dots, X_{m-1}, A_m$ . For the sake of simplicity, we use notations such as  $\bar{A}_p = X_{p-1} \cup A_p \cup X_p$ ,  $A_p^- = X_{p-1} \cup A_p$  and  $A_p^+ = A_p \cup X_p$  ( $p = 1, 2, \dots, m; X_0 = X_m = \emptyset$ ). A *cascade network* is a network such that two nodes are connected by a branch if and only if they belong to one and the same  $\bar{A}_p$ . Distance matrices on a cascade network  $G_{cs}$  take the form in Fig. 5.

Let us apply theorems 8 and 9 to a cascade network  $G_{cs}$ . For nodes  $i$  and  $j$  in the same set  $\bar{A}_p$  we have  $u(i, j) = |\bar{A}_p| - 2$ .

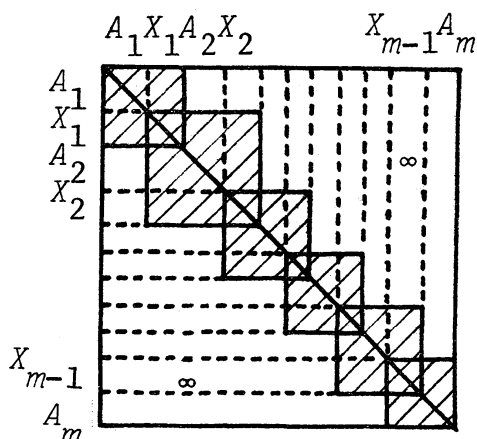


Fig. 5. A distance matrix corresponding to a cascade network.

For  $p$  and  $q$  such that  $1 \leq p < q \leq m$  we define  $\hat{y}(p, q)$  by  $|X_{\hat{y}(p, q)}| = \min(|X_p|, \dots, |X_{q-1}|)$ . Then, for nodes  $i \in A_p^-$  and  $j \in A_q^+$  we have  $u(i, j) = u(j, i) = \hat{y}(p, q)$ . Therefore, the complexity of any valid algorithm for a cascade network  $G_{st}$  is at least

$$\sum_{p=1}^m \bar{a}_p (\bar{a}_p - 1) (\bar{a}_p - 2) - \sum_{p=1}^{m-1} x_p (x_p - 1) (x_p - 2) + 2 \sum_{p < q} a_p^- a_q^+ x_{\hat{y}(p, q)}, \quad (4.1)$$

where  $a_p = |A_p|$ ,  $x_p = |X_p|$ ,  $\bar{a}_p = |\bar{A}_p|$ ,  $a_p^- = |A_p^-|$  and  $a_p^+ = |A_p^+|$ .

The following algorithm, proposed initially by us in [2], consists of only the above mentioned triple-operations, each occurring precisely once and, therefore, is optimal. This algorithm is a modified form of  $\Gamma_{13} \cdot \Gamma_2^B \cdot \Gamma_4^F$  in §3.6.

Let us denote the intersection of diagonal blocks with  $R(k)$  by  $R(k; D)$ , where  $R = I, II, III$  or  $IV$  (see Fig. 6). We define  $\Gamma_2^B(\text{in})$ ,  $\Gamma_4^F(\text{in})$  and  $\Gamma_{13}(\text{in})$  as operators obtained by replacing  $\langle R(k) \rangle^k$ 's in  $\Gamma_2^B$ ,  $\Gamma_4^F$  and  $\Gamma_{13}$ , respectively, with the corresponding  $\langle R(k; D) \rangle^k$ 's. So far as  $G_{cs}$  is concerned, it is readily seen that

$$\langle II(k) \rangle \equiv \langle II(k; D) \rangle, \quad \langle IV(k) \rangle \equiv \langle IV(k; D) \rangle,$$

because the values outside the diagonal blocks remains unchanged (i.e. remains to be  $\infty$ ) under these operators. Therefore, the

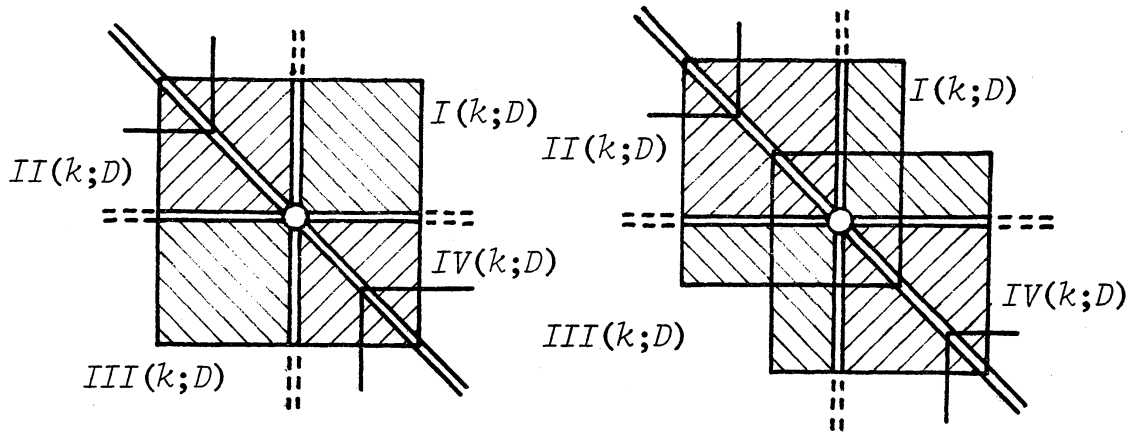


Fig. 6. Illustration of  $R(k; D)$ 's ( $R = I, II, III$  or  $IV$ ).

effect of  $\Gamma_2^B$  and  $\Gamma_4^F$  in  $\Gamma_{13} \cdot \Gamma_2^B \cdot \Gamma_4^F$  is equal to that of  $\Gamma_2^B(\text{in})$  and  $\Gamma_4^F(\text{in})$ , respectively. Note, also, that the values outside the diagonal blocks have no influence upon the diagonal blocks when we apply the algorithm  $\Gamma_{13}$ . Thus, the algorithm  $\Gamma_{13}(\text{in}) \cdot \Gamma_2^B(\text{in}) \cdot \Gamma_4^F(\text{in})$  computes the correct value of  $d_{ij}^\infty$ 's in the diagonal blocks.

Next, let  $i \in A_p^-$ ,  $j \in A_q^+$  and  $p < q$ . Then for any  $y(p, q)$  such that  $p \leq y(p, q) < q$ , we have

$$d_{ij}^\infty = \min_{k \in X_{y(p,q)}} (d_{ik}^\infty + d_{kj}^\infty), \tag{4.2}$$

because every elementary conductive paths from  $i$  to  $j$  passes  $X_{y(p,q)}$ . This fact implies that the values  $d_{ij}^\infty$ 's outside the diagonal blocks may be computed in an increasing order of  $q - p$  ( $i \in A_p^-$ ,  $j \in A_q^+$  or  $j \in A_p^-$ ,  $i \in A_q^+$ ;  $p < q$ ). Thus, we have the following algorithm, where we set  $y(p, q) = \hat{y}(p, q)$  in

order to make the algorithm optimal.

$$\left. \begin{aligned}
 \Gamma_{13}^{\text{(out)}} &\stackrel{\text{d}}{=} \Gamma_{13}^{m-1} \cdot \Gamma_{13}^{m-2} \cdot \dots \cdot \Gamma_{13}^2 \cdot \Gamma_{13}^1, \\
 \Gamma_{13}^r &\stackrel{\text{d}}{=} \prod_{q-p=r} \Gamma_{13}^{p,q} \quad (r = 1, 2, \dots, m-2, m-1), \\
 \Gamma_{13}^{p,q} &\stackrel{\text{d}}{=} \prod_{k \in X_{\hat{y}}(p,q)} \langle A_p^- \times A_q^+ \cup A_q^+ \times A_p^- \rangle \quad (p < q).
 \end{aligned} \right\} \quad (4.3)$$

As we have already mentioned, this algorithm  $\Gamma_{13}^{\text{(out)}} \cdot \Gamma_{13}^{\text{(in)}} \cdot \Gamma_{2}^B(\text{in}) \cdot \Gamma_{4}^F(\text{in})$  contains triple-operations of the number equal to (4.1) so that it is optimal.

If  $\alpha_1 = \alpha_2 = \dots = \alpha_m = \alpha$  and  $x_1 = x_2 = \dots = x_{m-1} = x$ , then the leading term (i.e. the terms of the order highest in  $\alpha$  and  $x$ ) of (4.1) is equal to

$$m\alpha^3 + (m^2+5m-6)\alpha^2x + (2m^2+6m-14)\alpha x^2 + (m^2+2m-7)x^3. \quad (4.4)$$

The leading term of the complexity of the algorithm in [14] or that in [15] for cascade networks is equal, respectively, to

$$(2m-1)\alpha^3 + (m^2+11m-15)\alpha^2x + (2m^2+18m-35)\alpha x^2 + (m^2+11m-23)x^3, \quad (4.5)$$

and

$$m\alpha^3 + (m^2+6m-7)\alpha^2x + (2m^2+10m-20)\alpha x^2 + (m^2+6m-14)x^3. \quad (4.6)$$

4.4. *Star networks.* A star network is another special case of tree-structured networks.

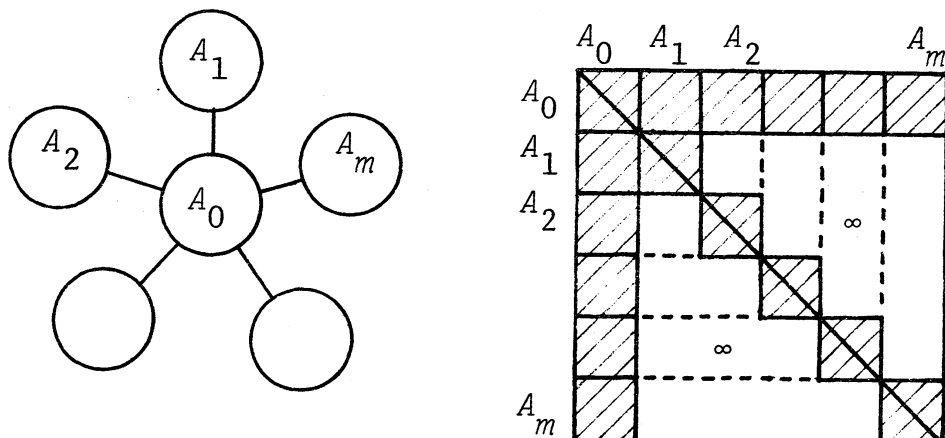


Fig. 7. A star network and its distance matrix.

If the set of nodes  $N$  of a network  $G_{st}$  is decomposed into disjoint subsets  $A_0, A_1, A_2, \dots, A_m$ , and if there is no branch connecting nodes  $i$  and  $j$  not belonging to the same subset  $\bar{A}_p = A_0 \cup A_p$ , we call the network a *star network*. An example of  $G_{st}$  with its distance matrix is shown in Fig. 7.

Theorem 10 shows that a triple-operation  $\langle i^k_j \rangle$  with pivot  $k \in A_p$  must appear at least once for every node-pair  $(i, j)$  in  $\bar{A}_p \times \bar{A}_p$  ( $p = 1, 2, \dots, m$ ), and a triple-operation  $\langle i^k_j \rangle$  with pivot  $k \in A_0$  must appear at least once for every node-pair  $(i, j)$  in  $N \times N$ . Thus the complexity of any valid algorithm for  $G_{st}$  is bounded from below by

$$a_0(n-1)(n-2) + \sum_{p=1}^m a_p(\bar{a}_p-1)(\bar{a}_p-2), \tag{4.7}$$

where  $a_p = |A_p|$  and  $\bar{a}_p = |\bar{A}_p|$  ( $p = 0, 1, 2, \dots, m$ ).

Let us apply the Warshall-Floyd algorithm for  $G_{st}$ , omitting all those operations ineffective by virtue of the structure of

$G_{st}$ . Since only the submatrix  $\bar{A}_p \times \bar{A}_p$  is transformed by  $\langle i^k j \rangle$  when  $k \in A_p$  ( $p = 1, 2, \dots, m$ ), we have

$$\langle N \times N \rangle^k \equiv \langle \bar{A}_p \times \bar{A}_p \rangle^k \quad \text{if } k \in A_p \quad (p = 1, 2, \dots, m).$$

Therefore, the algorithm

$$\Gamma_{st} \stackrel{d}{=} \left( \prod_{k \in A_p} \langle N \times N \rangle^k \right) \cdot \prod_{p=1}^m \left( \prod_{k \in A_p} \langle \bar{A}_p \times \bar{A}_p \rangle^k \right)$$

is valid for  $G_{st}$ . Moreover, the complexity of  $\Gamma_{st}$  is equal to (4.7). Thus  $\Gamma_{st}$  is optimal. If  $a_1 = a_2 = \dots = a_m = a$ , then the leading term of (4.7) is equal to

$$a_0 n^2 + ma(a_0 + a)^2. \quad (4.8)$$

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