

Multi-Source Multi-Sink Stochastic-Flow Networks Reliability under Time Constraints

M. R. Hassan¹ and H. Abdou^{1,2}

¹Department of Mathematics, Computer Science Branch, Mathematics Department, Aswan University, Aswan, Egypt; m_r_hassan73@yahoo.com

²Department Applied Natural Sciences, Qassim University, Buraydah, Saudi Arabia; haniabdou2000@yahoo.com

Abstract

Objectives: This study is centered on four issues related to the reliability evaluation in multi-source multi-sink networks. Each issue discusses the reliability evaluation under different condition. These conditions play an important role in determining the quickest paths used in transmitting data between source and sink nodes, with the condition that the transmission time of the quickest path does not surpass a predetermined upper bound T . **Methods/Statistical Analysis:** Proposed algorithms used in each issue designed based on approaches taken from previous literatures to evaluate the reliability. **Findings:** The reliability ($R_{d_{w_j}T}$) evaluated when each source transmits the demand d_{w_j} (the demand for resource w at sink node t_j) separately to the distinct sink, this is the first issue. The second issue deals with transmitting demands request by sink nodes from one source via a group of disjoint paths. In the case of transmitting demands through a gathering of joint paths, share one or more common arc, this is the third one. The last issue investigates the problem of sending demands requested by all sinks from all sources via joint paths. **Application/Improvements:** Examples are presented to illustrate how to evaluate the reliability of a multi-source multi-sink networking each case under time constraint.

Keywords: Joint and Disjoint Paths, Multi-Source Multi-Sink Stochastic-Flow Networks, Time Constraint, Quickest Path, System Reliability

1. Introduction

The quickest path problem is to obtain a routing path in a network with a minimum time to ship σ units of data from the source to the sink¹. In², the proposed method is targeted towards the situation where multi-commodities are conveyed through all disjointed minimal paths (MPs) in a network. In³, distributed algorithms are developed for the quickest path problem in any a synchronous communications networks. In⁴ the problem is supposed as a criteria path problem, allowing the use of a very efficient algorithm, which solves the quietest path problem for all possible values of the amount of data that has to be transmitted.

The system reliability of stochastic-flow networks under time constraint is defined as the probability of sending d units of data from the source to the sink through the network within T units of time, denoted by $R_{d,T}$ ⁵⁻⁷. The problem of determining the optimal routing policy with the highest system reliability discussed in⁸ and⁹. Network reliability has been evaluated in the case of sending units of data through a number of MPs simultaneously under both time and budget constraints¹⁰. Moreover, network reliability according to the spare routing was evaluated¹¹. In order to reduce transmission time, the problem of simultaneously transmitting data through multiple disjoint minimal paths was presented in¹².

*Author for correspondence

A multi-source multi-sink stochastic-flow network is an extension of the concept to multiple sources and sinks on the same network. Evaluating the system reliability of multi-source multi-sink stochastic flow networks has furthermore been addressed in¹³⁻¹⁶. In¹³, the optimal resource allocation problem subject to reliability maximization has been formulated and presented an algorithm to solve it. For more than one resource¹⁴, the optimal resource flow allocation problem has been studied and a GA was proposed to solve it. In¹⁵, the flow allocation problem subject to transportation cost was studied and solved using GAs. In¹⁶, the author modified and solved the formulation of the flow allocation problem subject to the probability of the capacity vector and transmission cost. Further, system reliability was evaluated by searching for the optimal lower boundary points

In this paper, we will extend the quickest path problem to multi-source multi-sink flow networks. The presented problem has been studied under the following cases:

1. Each source node (s_i) sends the specified demand d_{wj} (demand for resource w at sink node t_j) separately to each sink (t_j).
2. Each source node (s_i) sends the specified demands to all sink nodes ($t_j, j=1,2,\dots$, through different paths that do not share any common arcs (disjoint paths).
3. Each source node (s_i) sends the specified demands to all sink nodes ($t_j, j=1,2,\dots$, through joint paths that share some arcs.
4. All source nodes ($s_i, i=1,2,3,\dots$) sends multiple demands to all sink nodes ($t_j, j=1,2,\dots$) through joint paths, the general case, simultaneously transmitting.

The rest of the paper is organized as follows. Section 2 presents notations and assumptions. Section 3 presents Case A: transmitting demands separately. Section 4 describes Case B: transmitting demands through disjoint paths. Section 5 provides Case C: transmitting demands through joint paths. Case D: the general case, when transmitting multiple demands from all sources to all sinks given in Section 6. Section 7 offers our conclusions.

2. Notation and Assumptions

2.1 Notation

$G(A, N, M, S, T)$ a multi-source multi-sink stochastic-flow network.

$A = \{a_e \mid 1 \leq e \leq n\}$, set of arcs.

N set of nodes.

$M = \{M_1, M_2, \dots, M_n\}$, where M_e is the maximum capacity of each arc a_e .

$S = \{s_1, \dots, s_q\}$: set of source nodes.

$T = \{t_1, \dots, t_\theta\}$: set of sink nodes.

$D = \{d_{w,j} \mid 1 \leq w \leq m, 1 \leq j \leq \theta\}$, where $d_{w,j}$ is the demand for resource w at sink node t_j .

MP Minimal path,

$MP_{i,j,k}$ The k^{th} MP from s_i to t_j .

$MPS = \{MP_{i,j,k} \mid 1 \leq i \leq \sigma, 1 \leq j \leq \theta, 1 \leq k \leq k_{i,j}\}$: a set of all MPs, where $k_{i,j}$ represents the number of MPs from s_i to t_j .

L_i The lead time of arc

$L(MP_{i,j,k})$ The total lead time of the path($MP_{i,j,k}$)

$W^{wj}(a_l)$ is the consumed capacity of by commodity d_{wj} .

np Total number of MPs contained in MPS.

X Capacity vector defined as $X = (x_1, x_2, \dots, x_p, \dots, x_n)$.

$R_{d_w, T}$ The system reliability for the given demand d_{wj} under T.

2.2 Assumptions

1. The capacity of each arc a_e is an integer-valued random variable, which takes values $0 < 1 < 2 < \dots < M_e$ according to a given distribution.
2. The capacities of the arcs are statistically independent.
3. The flow along a path does not exceed its maximum capacity.

3. Case A: Transmitting Demands Separately from One Source to One Sink

The following subsections describe how to calculate i.e. the reliability of transmitting a single demand from the source node to the sink.

3.1 Definition of Lower Boundary Points for (d_{wj}, T)

If X is a minimal capacity vector such that the network can send d_{wj} units of data from the source to the sink within T units of time, then X is called a lower boundary point for (d_{wj}, T) .

3.2 Generate All Lower Boundary Points for (d_{wj}, T) .

In the following steps, for the k^{th} MP, $MP_{i,j,k}$ from s_i to t_j , $MP_{i,j,k} = \{a_1, \dots, a_n\}$, we will show how to find the minimal capacity vector $X^{i,j} = (x_1, x_2, \dots, x_n)$ such that the network sends d_{wj} units of data within T units of time from the source s_i to the sink t_j .

1. For each $MP_{i,j,k}$, determine the smallest integer v such that,

$$\sum_{i=1}^n \{1_i \mid a_i \in p_{j,j,k}\} + \lceil d_{jv} / v \rceil \leq T \dots (1)$$

2. If v , generate the system capacity vector $X^{i,j} = (x_1, x_2, \dots, x_n)$ for each $MP_{i,j,k}$ as follows:

$$x_e = \begin{cases} u \geq v & \text{if } a_i \in MP_{i,j,k} \\ 0 & \text{otherwise} \end{cases} \dots (2)$$

Where x_e is an element of $X^{i,j}$ and u is the minimal capacity of a_i .

3.3 Evaluation of $R_{d_{wj}, T}$

If $X_1^{i,j}, X_2^{i,j}, \dots, X_q^{i,j}$ are the collection of all (d_{wj}, T) - $MP_{i,j}$, and then the system reliability $R_{d_{wj}, T}$ is defined as follows:

$$R_{(d_{wj}, T)} = \Pr\{\bigcup_{u=1}^q \{Y \mid Y \geq X_u^{i,j}\}\} \dots (3)$$

Where $\Pr\{Y\} = \Pr\{y_1\} \cdot \Pr\{y_2\} \cdot \dots \cdot \Pr\{y_n\}$.

Several methods¹⁶⁻²⁰ can be used to evaluate (3), in this paper, we will use to evaluate $R_{d_{wj}, T}$.

3.4 Illustrative Example

As an example, we consider the network in Figure 1, which has two source and two sink nodes. The arcs are numbered from a_1 to a_{14} ; their capacities, corresponding probabilities and lead-time of each arc taken from^{15,16}.

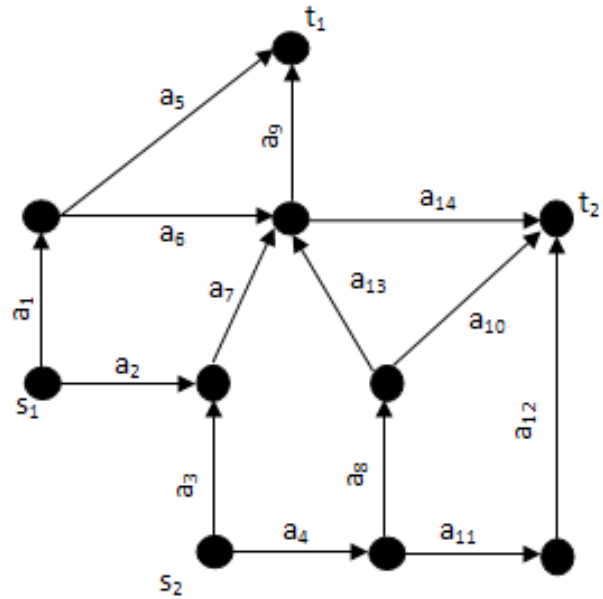


Figure 1. Two-source two-sink computer network.

In the following steps, we will show how to calculate $R_{d_{11}, T}$, i.e., the reliability from the source node s_1 to the sink node t_1 , where $d_{11} = 11$ and $T = 9$, i.e. evaluate $R_{11,9}$.

There are three MPs from s_1 to t_1 : $MP_{1,1,1} = \{a_1, a_5\}$, $MP_{1,1,2} = \{a_1, a_6, a_9\}$, $MP_{1,1,3} = \{a_2, a_7, a_9\}$.

Step 1: (1.1) the lead-time of $MP_{1,1,1} = \{a_1, a_5\}$ is $l_1 + l_5 = 3$. Then $v=2$ is the smallest integer such that

$$\left(3 + \left\lceil \frac{1}{2} \right\rceil\right) \leq 9.$$

(1.2) the maximal capacity of $MP_{1,1,1}$ is 8. Hence, $x_1 = x_5 = 2$ and $x_i = 0$ for others. So we obtain $X_1 = (2, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0)$.

Step 2: (1.1) the lead-time of $MP_{1,1,2} = \{a_1, a_6, a_9\}$ is $l_1 + l_6 + l_9 = 7$. Then $v=6$ is the smallest integer

$$\text{Such that } \left(7 + \left\lceil \frac{1}{6} \right\rceil\right) \leq 9.$$

(1.2) the maximal capacity of $MP_{1,1,2}$ is 10. Hence, $x_1 + x_6 + x_9 = 6$ and $x_i = 0$ for others. So we obtain $X_2 = (6, 0, 0, 0, 0, 6, 0, 0, 6, 0, 0, 0, 0, 0)$.

Step 3: (1.1) the lead-time of $MP_{1,1,3} = \{a_2, a_7, a_9\}$ is $l_2 + l_7 + l_9 = 6$. Then $v=4$ is the smallest integer such that

$$(6 + \left\lceil \frac{1}{4} \right\rceil) \leq 9.$$

(1.2) the maximal capacity of $MP_{1,1,3}$ is 10. Hence, $x_2 + x_7 + x_9 = 4$ and $x_i = 0$ for others. So we obtain $X_3 = (0, 4, 0, 0, 0, 0, 4, 0, 4, 0, 0, 0, 0, 0)$.

Hence $R_{11,9} = 0.999957$ using Eq (3). Table 1 summarizes the values of $R_{d_{w_j}, T}$ for different values of d_{w_j} .

Table 1. Values of $R_{d_{w_j}, T}$ for different values of d_{w_j}

(s_i, t_j)	MP_{ijk}	$R_{d_{w_j}, T}$	The value of $R_{d_{w_j}, T}$
(1,2)	$MP_{1,2,1} = \{a_1, a_6, a_{14}\}$ $MP_{1,2,2} = \{a_2, a_7, a_{14}\}$	$R_{12,9}$	0.995772
(2,1)	$MP_{2,1,1} = \{a_3, a_7, a_9\}$ $MP_{2,1,2} = \{a_4, a_8, a_9, a_{13}\}$	$R_{7,9}$	0.993354
(2,2)	$MP_{2,2,1} = \{a_3, a_7, a_{14}\}$ $MP_{2,2,2} = \{a_4, a_8, a_{13}, a_{14}\}$ $MP_{2,2,3} = \{a_4, a_8, a_{10}\}$ $MP_{2,2,4} = \{a_4, a_1, a_{12}\}$	$R_{10,9}$	0.999881

4. Case B: Transmitting Demands through Disjoint Paths

We study how to calculate $R_{D_{w_j}^u, T}$, the reliability from the source node s_1 to the sink nodes t_1 and t_2 , where $D_{w_j}^u = (d_{11}, d_{12})$. There are five MPs from s_1 to both t_1 and t_2 shown in Table 2-3. In this case, the following constraint for each bandwidth should be satisfied when is sent through.

$$b_k^{w_j} \leq B_k \tag{4}$$

Table 2. The consumed capacity

Arc	$W^{11}(a_i)$	$W^{12}(a_i)$	Arc	$W^{11}(a_i)$	$W^{12}(a_i)$
a_1	1	2	a_8	1	2
a_2	1	2	a_9	1	2

a_3	1	2	a_{10}	1	2
a_4	1	2	a_{11}	1	2
a_5	1	2	a_{12}	1	2
a_6	1	2	a_{13}	1	2
a_7	1	2	a_{14}	1	2

Table 3. The data of each MP and B_k values

k	MP_{ijk}	$L(MP_{ijk})$	B_k
1	$\{a_1, a_3\}$	3	8
2	$\{a_1, a_6, a_9\}$	7	10
3	$\{a_2, a_7, a_9\}$	6	10
4	$\{a_1, a_6, a_{14}\}$	6	10
5	$\{a_2, a_7, a_{14}\}$	5	10

Where c is the upper bound for bandwidth of, it is given by:

$$B_k = \min_{a_i \in MP_{ijk}} \{c(a_i)\} \tag{5}$$

The $b_k^{w_j}$ bandwidth is evaluated by the following equation:

$$b_k^{w_j} = \frac{\max_{a_i \in MP_{ijk}} \{W^{w_j}(a_i)\} \times d_{w_j}}{T - L(MP_{ijk})} \tag{6}$$

Finally, each x_h in the capacity vector $X, X = ()$, is constructed by eq. (7), (2).

$$x_h = \sum_{k=1}^{np} \sum_{j=1}^{m, \theta} \rho_k b_k^{w_j}, \text{ where } \rho_k \begin{cases} 1, & \text{if } a_h \in MP_{ijk} \\ 0 & \text{if } a_h \notin MP_{ijk} \end{cases} \tag{7}$$

The following algorithm is used to evaluate the reliability for Case B.

Algorithm B
Begin

- B.1. For each arc a_i ; Read p_i , and.
- B.2. Determine the source node s_i and the sink nodes t_1, t_2 .
- B.3. Read T , and d_{w_j} .
- B.4. For $k=1$ to np do
 - B.4.1. Calculate $b_k^{w_j}$ and according to Eq. (5) and (6) respectively.
 - B.4.2. End do
- B.5. Determine the set of disjoint paths.

B.6. Construct the capacity vector X using Eq. (7).

B.7. Evaluate using Eq (3).

End.

Given = (10, 10) and T = 11, with the consumed capacity shown in Table 2. Table 4,5 summarizes the values of, the set of disjoint paths and respectively.

Table 4. The disjoint MPS

Disjoint Pair	Commonly-used arcs
$MP_{111} \cap MP_{113}$	φ
$MP_{111} \cap MP_{122}$	φ
$MP_{112} \cap MP_{122}$	φ
$MP_{113} \cap MP_{121}$	φ

Table 5. w_j Values

j	MP_{ijk}	b_k^{11}	b_k^{12}	B_k
1	$\{a_1, a_5\}$	2	3	8
2	$\{a_1, a_6, a_9\}$	3	5	10
3	$\{a_2, a_7, a_9\}$	2	4	10
4	$\{a_1, a_6, a_{14}\}$	2	4	10
5	$\{a_2, a_7, a_{14}\}$	2	3	10

Finally, Table 6 summarizes the candidate vectors. The corresponding reliability is =0.999988. The reliability values for other sources and sinks are shown in Table 7.

Table 6. The candidate capacity vectors

(b^{11}, b^{12})	$(x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8, x_9, x_{10}, x_{11}, x_{12}, x_{13}, x_{14})$	QP candidate
$(b_1^{11}, b_3^{12})=(2,4)$	(2,4,0,0,2,0,4,0,4,0,0,0,0,0)	X_1
$(b_3^{11}, b_1^{12})=(2,3)$	(3,2,0,0,3,0,2,0,2,0,0,0,0,0)	X_2
$(b_1^{11}, b_5^{12})=(2,3)$	(2,3,0,0,2,0,3,0,0,0,0,0,0,3)	X_3
$(b_5^{11}, b_1^{12})=(2,3)$	(3,2,0,0,3,0,2,0,0,0,0,0,0,2)	X_4
$(b_5^{11}, b_2^{12})=(2,5)$	(5,2,0,0,0,5,2,0,5,0,0,0,0,2)	X_5
$(b_5^{11}, b_5^{12})=(3,3)$	(3,3,0,0,3,0,3,0,3,0,0,0,0,3)	X_6
$(b_4^{11}, b_5^{12})=(2,4)$	(2,4,0,0,0,2,4,0,4,0,0,0,0,2)	X_7
$(b_3^{11}, b_4^{12})=(2,4)$	(4,2,0,0,0,4,2,0,2,0,0,0,0,4)	X_8

Table 7. Values of $R_{D^2, T}$ for different values of d_{wj}

Between different sources and sinks	MP_{ijk}	$R_{D^2, T}$
$((1,2) \rightarrow 1)$	$MP_{1,2,1^1} = \{a_1, a_5\}$ $MP_{1,2,1^2} = \{a_1, a_6, a_9\}$ $MP_{1,2,1^3} = \{a_2, a_7, a_9\}$ $MP_{1,2,1^4} = \{a_3, a_7, a_9\}$ $MP_{1,2,1^5} = \{a_4, a_8, a_9, a_{13}\}$	0.999990
$(2 \rightarrow (1,2))$	$MP_{2,1,2^1} = \{a_3, a_7, a_9\}$ $MP_{2,1,2^2} = \{a_4, a_8, a_9, a_{13}\}$ $MP_{2,1,2^3} = \{a_3, a_7, a_{14}\}$ $MP_{2,1,2^4} = \{a_4, a_8, a_{13}, a_{14}\}$ $MP_{2,1,2^5} = \{a_4, a_8, a_{10}\}$ $MP_{2,1,2^6} = \{a_4, a_{11}, a_{12}\}$	0.999986
$((1,2) \rightarrow 2)$	$MP_{1,2,2^1} = \{a_1, a_6, a_{14}\}$ $MP_{1,2,2^2} = \{a_2, a_7, a_{14}\}$ $MP_{1,2,2^3} = \{a_3, a_7, a_{14}\}$ $MP_{1,2,2^4} = \{a_4, a_8, a_{13}, a_{14}\}$ $MP_{1,2,2^5} = \{a_4, a_8, a_{10}\}$ $MP_{1,2,2^6} = \{a_4, a_{11}, a_{12}\}$	0.999995

5. Case C: Transmitting Demands Via Joint Paths from One Source to all Sinks

We study how to calculate $R_{D^2, T}$, the reliability from

the source node s_1 to the sink node t_1 and t_2 , where. There are five MPs from s_1 to both t_1 and t_2 shown in Table 8 and 9. In this case, the following constraint for each bandwidth should satisfy Eqs. (4), (5), (6) and

$$\sum_{j=1}^m \sum_{i=1}^u \rho_j b_j^i \leq c(a^*) \tag{8}$$

Table 8. The consumed capacity

Arc	$W^{11}(a_1)$	$W^{12}(a_1)$	Arc	$W^{11}(a_1)$	$W^{12}(a_1)$
a_1	1	2	a_8	1	2
a_2	1	2	a_9	1	2
a_3	1	2	a_{10}	1	2
a_4	1	2	a_{11}	1	2
a_5	1	2	a_{12}	1	2
a_6	1	2	a_{13}	1	2
a_7	1	2	a_{14}	1	2

Table 9. The data of each MP and B_k values

k	MP_{ijk}	$L(MP_{ijk})$	B_k
1	$\{a_1, a_5\}$	3	8

2	{a ₁ ,a ₆ ,a ₉ }	7	10
3	{a ₂ ,a ₇ ,a ₉ }	6	10
4	{a ₁ ,a ₆ ,a ₁₄ }	6	10
5	{a ₂ ,a ₇ ,a ₁₄ }	5	10

where a^{*} is a commonly-used arc of two or more MPs.

Given D²= (10, 10) and T = 11, with the consumed capacity shown in Tables 10,11 summarizes the values of , the set of joint paths and respectively.

Table 10. The joint MPS

Joint Pair	Commonly-used arcs (a [*])
MP ₁₁₁ , MP ₁₁₂ , MP ₁₂₁	{a ₁ }
MP ₁₁₂ , MP ₁₁₃	{a ₉ }
MP ₁₁₂ , MP ₁₂₁	{a ₆ }
MP ₁₁₃ , MP ₁₂₂	{a ₇ }
MP ₁₂₁ , MP ₁₂₂	{a ₁₄ }

Table 11. b_k^{wj} values

j	MP _{ijk}	b _j ¹	b _j ²	C(a [*])
1	{a ₁ ,a ₅ }	2	3	8
2	{a ₁ ,a ₆ ,a ₉ }	3	5	10
3	{a ₂ ,a ₇ ,a ₉ }	2	4	10
4	{a ₁ ,a ₆ ,a ₁₄ }	2	4	10
5	{a ₂ ,a ₇ ,a ₁₄ }	2	3	10

Finally, each x_i in the capacity vector X, X=(x₁,x₂,x₃,x₄,x₅,x₆,x₇,x₈,x₉,x₁₀,x₁₁,x₁₂,x₁₃,x₁₄), is constructed by Eq. (7). Table 12 summarizes the candidate vectors, the corresponding R_{D²,1} = 0.999988 the reliability value from s2 to both t1 and t2 shown in Table 13.

Table 12. The candidate capacity vectors

(b ¹¹ ,b ¹²)	(x ₁ ,x ₂ ,x ₃ ,x ₄ ,x ₅ ,x ₆ ,x ₇ ,x ₈ ,x ₉ ,x ₁₀ ,x ₁₁ ,x ₁₂ ,x ₁₃ ,x ₁₄)	QP candidate
(b ₁ ¹¹ ,b ₂ ¹²)=(2,5)	(7,0,0,0,2,5,0,0,5,0,0,0,0,0)	x ₁
(b ₂ ¹¹ ,b ₁ ¹²)=(3,3)	(6,0,0,0,3,3,0,0,3,0,0,0,0,0)	x ₂
(b ₁ ¹¹ ,b ₄ ¹²)=(2,4)	(6,0,0,0,2,4,0,0,0,0,0,0,0,4)	x ₃

(b ₄ ¹¹ ,b ₁ ¹²)=(2,3)	(5,0,0,0,3,2,0,0,0,0,0,0,0,2)	x ₄
(b ₂ ¹¹ ,b ₃ ¹²)=(3,4)	(3,4,0,0,0,3,4,0,7,0,0,0,0,0)	x ₅
(b ₃ ¹¹ ,b ₂ ¹²)=(2,5)	(5,2,0,0,0,5,2,0,7,0,0,0,0,0)	x ₆
(b ₂ ¹¹ ,b ₄ ¹²)=(3,4)	(7,0,0,0,0,7,0,0,3,0,0,0,0,4)	x ₇
(b ₄ ¹¹ ,b ₂ ¹²)=(2,5)	(7,0,0,0,0,7,0,0,5,0,0,0,0,2)	x ₈
(b ₃ ¹¹ ,b ₅ ¹²)=(2,3)	(0,5,0,0,0,0,5,0,2,0,0,0,0,3)	x ₉
(b ₅ ¹¹ ,b ₃ ¹²)=(2,4)	(0,6,0,0,0,0,6,0,4,0,0,0,0,2)	x ₁₀
(b ₄ ¹¹ ,b ₅ ¹²)=(2,4)	(2,3,0,0,0,2,3,0,0,0,0,0,0,5)	x ₁₁
(b ₅ ¹¹ ,b ₄ ¹²)=(2,4)	(4,2,0,0,0,4,2,0,0,0,0,0,0,6)	x ₁₂

Table 13. Values of R_{D²,T} for different values of d_{wj}

Between different sources and sinks	MP _{ij,k}	R _{D²,T}	The value of R _{D²,T}
(2→(1,2))	MP _{2,1,2,1} ={a ₄ ,a ₈ ,a ₉ ,a ₁₃ } MP _{2,1,2,2} ={a ₃ ,a ₇ ,a ₁₄ } MP _{2,1,2,3} ={a ₄ ,a ₈ ,a ₁₃ ,a ₁₄ } MP _{2,1,2,4} ={a ₄ ,a ₈ ,a ₁₀ } MP _{2,1,2,5} ={a ₄ ,a ₁₁ ,a ₁₂ }	R _{D²,11}	0.999968

6. Case D: Transmitting Multiple Demands Via Joint Paths from all Sources to all Sinks

The system reliability R_{D,T} of the multi-source multi-sink flow network can be calculated by using the inclusion-exclusion rule according to the generated set of all lower boundary points for (D, T). Given D⁴= (5, 5, 5, 5) and T = 11, with the consumed capacity W¹¹(a₁)= W¹²(a₁)=1,W²¹(a₁)=W²²(a₁)=2.. Table 14-16 summarizes the values of, the set of joint paths and, and the candidate capacity vectors respectively. The corresponding reliability R_{D²,11} = 0.999998. In addition, using different consumed capacity W¹¹ (a₁)= W¹² (a₁)=W²¹ (a₁)=W²² (a₁)=1. Table 17-19 summarizes the values of, the set of joint paths and, and the candidate capacity vectors respectively. The corresponding reliability R_{D²,11} = 0.999998.

Table 14. The data of each MP and B_k values

k	MP _{ijk}	L(MP _{ijk})	B _k
1	{a ₁ ,a ₅ }	3	8
2	{a ₁ ,a ₆ ,a ₉ }	7	10
3	{a ₂ ,a ₇ ,a ₉ }	6	10
4	{a ₁ ,a ₆ ,a ₁₄ }	6	10
5	{a ₂ ,a ₇ ,a ₁₄ }	5	10
6	{a ₃ ,a ₇ ,a ₉ }	8	10
7	{a ₃ ,a ₇ ,a ₁₄ }	7	10
8	{a ₄ ,a ₈ ,a ₉ ,a ₁₃ }	8	12
9	{a ₄ ,a ₈ ,a ₁₃ ,a ₁₄ }	7	12
10	{a ₄ ,a ₈ ,a ₁₀ }	5	8
11	{a ₄ ,a ₁₁ ,a ₁₂ }	7	10

Table 15. b_k^{wj} values

j	MP _{ijk}	b _j ¹¹	b _j ¹²	b _j ²¹	b _j ²²
1	{a ₁ ,a ₅ }	1	2	1	2
2	{a ₁ ,a ₆ ,a ₉ }	2	3	2	3
3	{a ₂ ,a ₇ ,a ₉ }	1	2	1	2
4	{a ₁ ,a ₆ ,a ₁₄ }	1	2	1	2
5	{a ₂ ,a ₇ ,a ₁₄ }	1	2	1	2
6	{a ₃ ,a ₇ ,a ₉ }	2	4	2	4
7	{a ₃ ,a ₇ ,a ₁₄ }	2	3	2	3
8	{a ₄ ,a ₈ ,a ₉ ,a ₁₃ }	2	4	2	4
9	{a ₄ ,a ₈ ,a ₁₃ ,a ₁₄ }	2	3	2	3
10	{a ₄ ,a ₈ ,a ₁₀ }	1	2	1	2
11	{a ₄ ,a ₁₁ ,a ₁₂ }	2	3	2	3

Table 16. The candidate capacity vectors

(b ¹¹ , b ¹² , b ²¹ , b ²²)	(x ₁ ,x ₂ ,x ₃ ,x ₄ ,x ₅ ,x ₆ ,x ₇ ,x ₈ ,x ₉ ,x ₁₀ , x ₁₁ ,x ₁₂ ,x ₁₃ ,x ₁₄)	QP candidate
a5 {a ₁ ,a ₅ }	(7,0,0,0,2,5,0,0,5,0,0,0,0)	x ₁
a10 {a ₄ ,a ₈ ,a ₁₀ }	(6,0,0,0,3,3,0,0,3,0,0,0,0)	x ₂
a11 {a ₄ ,a ₁₁ ,a ₁₂ }	(6,0,0,0,2,4,0,0,0,0,0,0,4)	x ₃
a12 {a ₄ ,a ₁₁ ,a ₁₂ }	(5,0,0,0,3,2,0,0,0,0,0,0,2)	x ₄

Table 17. The data of each MP and B_k values

k	MP _{ijk}	L(MP _{ijk})	B _k
1	{a ₁ ,a ₅ }	3	8
2	{a ₁ ,a ₆ ,a ₉ }	7	10

3	{a ₂ ,a ₇ ,a ₉ }	6	10
4	{a ₁ ,a ₆ ,a ₁₄ }	6	10
5	{a ₂ ,a ₇ ,a ₁₄ }	5	10
6	{a ₃ ,a ₇ ,a ₉ }	8	10
7	{a ₃ ,a ₇ ,a ₁₄ }	7	10
8	{a ₄ ,a ₈ ,a ₉ ,a ₁₃ }	8	12
9	{a ₄ ,a ₈ ,a ₁₃ ,a ₁₄ }	7	12
10	{a ₄ ,a ₈ ,a ₁₀ }	5	8
11	{a ₄ ,a ₁₁ ,a ₁₂ }	7	10

Table 18. b_k^{wj} values

j	MP _{ijk}	b _j ¹	b _j ²	b _j ³	b _j ⁴
1	{a ₁ ,a ₅ }	1	1	1	1
2	{a ₁ ,a ₆ ,a ₉ }	2	2	2	2
3	{a ₂ ,a ₇ ,a ₉ }	1	1	1	1
4	{a ₁ ,a ₆ ,a ₁₄ }	1	1	1	1
5	{a ₂ ,a ₇ ,a ₁₄ }	1	1	1	1
6	{a ₃ ,a ₇ ,a ₉ }	2	2	2	2
7	{a ₃ ,a ₇ ,a ₁₄ }	2	2	2	2
8	{a ₄ ,a ₈ ,a ₉ ,a ₁₃ }	2	2	2	2
9	{a ₄ ,a ₈ ,a ₁₃ ,a ₁₄ }	2	2	2	2
10	{a ₄ ,a ₈ ,a ₁₀ }	1	1	1	1
11	{a ₄ ,a ₁₁ ,a ₁₂ }	2	2	2	2

Table 19. The candidate capacity vectors

(b ¹ , b ² , b ³ , b ⁴)	(x ₁ ,x ₂ ,x ₃ ,x ₄ ,x ₅ ,x ₆ ,x ₇ ,x ₈ ,x ₉ ,x ₁₀ , x ₁₁ ,x ₁₂ ,x ₁₃ ,x ₁₄)	QP candidate
a2{a ₂ ,a ₇ ,a ₉ } and {a ₂ ,a ₇ ,a ₁₄ }	(0,8,0,0,0,0,8,0,4,0,0,0,4)	x ₁
a5{a ₁ ,a ₅ }	(0,4,0,0,4,0,0,0,0,0,0,0,0)	x ₂
a10{a ₄ ,a ₈ ,a ₁₀ }	(0,0,0,4,0,0,0,4,0,4,0,0,0)	x ₃
a11{a ₄ ,a ₁₁ ,a ₁₂ }	(0,0,0,8,0,0,0,0,0,0,8,8,0)	x ₄
a12{a ₄ ,a ₁₁ ,a ₁₂ }	(0,0,0,8,0,0,0,0,0,0,8,8,0)	x ₅
a2{a ₂ ,a ₇ ,a ₉ } and {a ₂ ,a ₇ ,a ₁₄ }	(0,8,0,0,0,0,8,0,4,0,0,0,4)	x ₆

7. Conclusions

The study is successfully evaluated the reliability in multi-source multi-sink stochastic flow networks in different

situations. Given the network information (arcs capacities, probabilities, lead times), required demand single or multiple based on the source and sink nodes, and the time constraints T. Situation A, transmitting with no restriction on determining the relationship between the group of paths, joint or disjoint. Situation B, transmitting the required demands *via* disjoint paths situation C, transmitting demands through a group of joint paths, share one or more common arc. Finally, the general case when all sources send the required demands to the sink nodes via joint paths, evaluating, of a multi-source multi-sink flow network.

8. References

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