

# On Deadlocking in Queueing Networks

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## Abstract

In this paper a deadlock in open restricted queueing networks is investigated. A method to detect when deadlock occurs in simulations of these networks is presented, using a state digraph. Markov models of five deadlocking queueing networks are given, and their properties on expected time to deadlock explored. These networks are a one node single server, one node multi server, two node single server without loops, two node multi server without loops, and two node single server with loops. These properties are compared to results obtained using the simulation model. Finally a bound on the time to deadlock of the two node single server with loops network is derived.

## 1 Introduction

Restricted open queueing networks that can experience deadlock are under-discussed in the literature. This paper addresses this by investigating deadlock properties in these queueing networks. A method of detecting deadlock in discrete event simulations of queueing networks is presented, and markovian models of five open restricted queueing networks are analysed, and the properties of their time to deadlock investigated. Finally a bound on the time to deadlock is given for one of these queueing networks.

Central to the study of deadlock in restricted queueing networks is the concept of blocking. Given two queues in tandem such that customers leaving the first service station enter the second, and the second queue has limited queueing capacity. If the second queue is full, and a customer finishes service at the first queue, that customer cannot join the next queue due to lack of capacity. This customer remains with the server, blocking other customers from beginning service with that server, until space becomes available at the second queue. This is referred to as blocking.

Throughout this paper service centers will be referred to as nodes, and an open unrestricted queueing network will use the following notation for the  $i$ th node:

- $\Lambda_i$  denotes the external arrival rate.
- $\mu_i$  denotes the service rate.
- $c_i$  denotes the number of parallel servers.
- $n_i$  denotes the queueing capacity.

- $r_{ij}$  denotes the routing probability from node  $i$  to node  $j$  upon completion of service at node  $i$ .

Exponential service times and Poisson arrivals are assumed.

For the purposes of this paper, deadlock is defined as follows.

**Definition 1.** *When a simulation is in a situation where at least one service station, despite having arrivals, ceases to begin or finish any more services due to recursive upstream blocking, the system is said to be in deadlock.*

Figure 1 shows a three node queueing network in a deadlocked state. The customers occupying servers  $B_1$  and  $B_2$  are blocked from entering the top node, while the customer occupying server  $A_1$  is blocked from entering the middle node. Due to mutual blocking, these customers are preventing any more natural movement in these two nodes. Note however that on part of the network need be in deadlock, as the bottom node is free to continue services as normal.

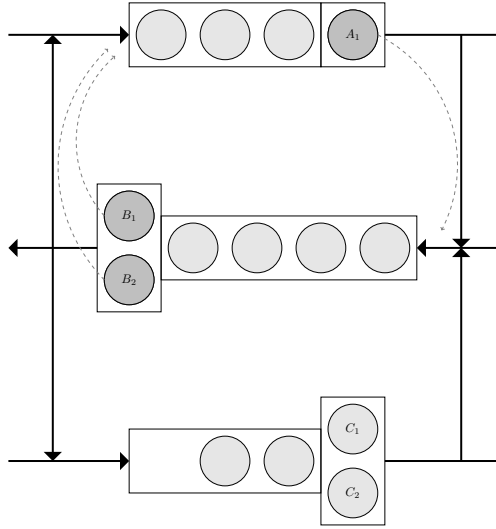


Figure 1: A three node queueing network in deadlock.

This paper is structured as follows: Section 2 discusses the exsisting literature on deadlock and deadlock strategies. Section 3 presents a method of detecting deadlock in discrete event simulations of queueing networks. Section 4 briefly describes the discrete event simulation model used to obtain the results in this paper. Section 5 presents Markov models of five deadlocking queueing networks, derives expected time to deadlock, and compares with resutls obtain through the simulation model. Finally Section 6 derives a bound for the expected time to deadlock for one of the queueing networks discussed.

## 2 Literature Review

Restricted queueing networks that exhibit blocking are well discussed in the literature [1, 2, 8, 10, 11, 13, 16]. Discussion on restricted queueing networks with feedback loops, that may exhibit deadlock, are sparse

however.

General deadlock situations that are not specific to queueing networks are discussed in [5]. Conditions for this type of deadlock, also referred to as deadly embraces, to potentially occur are given:

- Mutual exclusion: Tasks have exclusive control over resources.
- Wait for: Tasks do not release resources while waiting for other resources.
- No preemption: Resources cannot be removed until they have been used to completion.
- Circular wait: A circular chain of tasks exists, where each task requests a resource from another task in the chain.

In open restricted queueing networks the mutual exclusion condition is satisfied as customers cannot share servers; the wait for condition is satisfied due to the blocking rules defined previously; the no preemption condition is satisfied in networks that have no or non-preemptive priority (this report will only look at networks with no priority); and the circular wait condition is satisfied if the queueing network contains a cycle where all nodes have limited queueing capacity, that is feedback loops.

In general there are three strategies for dealing with deadlock [7, 9]:

- Prevention, in which the system cannot possibly deadlock in the first place.
- Avoidance, in which decisions are made as time unfolds to avoid reaching deadlock.
- Detection and recovery.

## 2.1 Deadlock Prevention

Deadlock prevention has been discussed in queueing networks. For closed networks of  $K$  customers with only one class of customer, [12] proves the following condition to ensure no deadlock: for each minimum cycle  $C$ ,  $K < \sum_{j \in C} B_j$ , the total number of customers cannot exceed the total queueing capacity of each minimum subcycle of the network. The paper also presents algorithms for finding the minimum queueing space required to ensure deadlock never occurs, for closed cactus networks, where no two cycles have more than one node in common. This result is extended to multiple classes of customer in [14], with more restrictions such as single servers and each class having the same service time distribution. Here an integer linear program is formulated to find the minimum queueing space assignment that prevents deadlock. The literature does not discuss deadlock properties in open restricted queueing networks.

## 2.2 Deadlock Avoidance

There are algorithms discussed in the literature for the dynamic avoidance of deadlock. In the Banker's Algorithm [6, 9], unsafe states, those that will lead to deadlock, are avoided by ensuring actions leading to these states are not carried out.

### 2.3 Deadlock Detection & Recovery

General deadlock detection in systems unspecific to queueing networks are discussed in [5]. A popular method of detecting general deadlock is the use of wait-for graphs, state-graphs and their variants [3, 4, 5, 7]. These wait-for graphs, keep track of all circular wait relations between tasks.

In [5] dynamic state-graphs are defined with resources as vertices and requests as edges. For scenarios where there is only one type of each resource, deadlock arises if and only if the state-graph contains a cycle. In [4] the vertices and edges of the state graph are given labels in relation to a reference node. Using these labels *simple bounded circuits* are defined whose existence within the state graph is sufficient to detect deadlock.

## 3 Detecting Deadlock

In order to detect when deadlock has occurred in a queueing network simulation, the state digraph is used, defined below.

**Definition 2.** *The state digraph  $D(t)$  of a queueing network defines that network's state at any time  $t$ . Vertices of the state digraph correspond to servers of the network. A directed edge denotes a blockage relationship in the following manner: if a customer at the  $k$ th server of node  $i$  is blocked from entering node  $j$ , then there are directed edges from the vertex corresponding to node  $i$ 's  $k$ th server to every vertex corresponding to the servers of node  $j$ .*

To illustrate this concept Figure 2 and Figure 3 show a three node network in and out of deadlock, and the corresponding state digraph in each case.

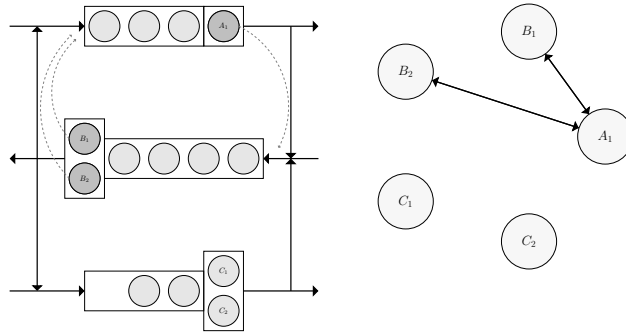


Figure 2: A three node queueing network in deadlock, with state digraph.

Consider one weakly connected component  $G(t)$  of  $D(t)$ . Consider the node  $X_a \in G(t)$ . Some observations:

- Consider the node  $X_a \in G(t)$ . If  $X_a$  is unoccupied, then  $X_a$  has no incident edges.
- Consider the case when  $X_a$  is occupied by individual  $a$ , whose next destination is node  $j$ . Then  $X_a$ 's direct successors are the servers occupied by individuals who are blocked or in service at node  $j$ .

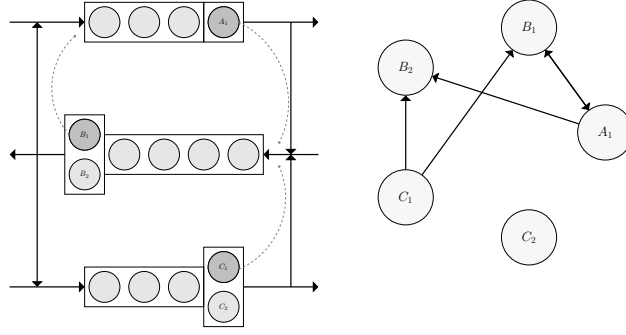


Figure 3: A three node queueing network not in deadlock, with state digraph.

- It can be interpreted that all  $X_a$ 's descendants are the servers whose occupants are directly or indirectly blocking  $a$ , and interpret all  $X_a$ 's ancestors as those servers whose individuals who are being blocked directly or indirectly by  $a$ .
- Note that the only possibilities for  $\deg^{\text{out}}(X_a)$  are 0 or  $c_j$ . If  $\deg^{\text{out}}(X_a) = c_j$  then  $a$  is blocked by all its direct successors. The only other situation is that  $a$  is not blocked, and  $X_a \in G(t)$  because  $a$  is in service at  $X_a$  and blocking other individuals, in which case  $\deg^{\text{out}}(X_a) = 0$ .
- It is clear that if all of  $X_a$ 's descendants are occupied by blocked individuals, then the system is deadlocked at time  $t$ .
- By definition all of  $X_a$ 's ancestors are occupied by blocked individuals.

The following results detect deadlock for open restricted queueing networks.

**Theorem 1.** *A deadlocked state arises at time  $t$  if and only if  $D(t)$  contains a knot.*

*Proof.* Consider one weakly connected component  $G(t)$  of  $D(t)$  at time  $t$ . All vertices of  $G(t)$  are either descendants of another vertex and so are occupied by an individual who is blocking someone; or are ancestors of another vertex, and so are occupied by someone who is blocked.

Assume that  $G(t)$  contains a vertex  $X$  such that  $\deg^{\text{out}}(X) = 0$ , and there is a path from every other non-sink vertex to  $X$ . This implies that  $X$ 's occupant is not blocked and is a descendant of another vertex. Therefore  $Q$  is not deadlocked as there does not exist a vertex whose descendants are all blocked.

Now assume that we have deadlock. For a vertex  $X$  who is deadlocked, all descendants of  $X$  are occupied by individuals who are blocked, and so must have out-degrees greater than 0. And so there is no path from  $X$  to a vertex with out-degree of 0.

□

**Proposition 1.** *For queueing networks:*

- with one node

- *with two nodes, each with two or fewer parallel servers*
- *with a finite amount of nodes, each with a single server*

*a deadlocked state arises if and only if there exists a weakly connected component without a sink node.*

*Proof.* Consider a one node queueing network.

If there is deadlock, then all servers are occupied by blocked individuals, and so all servers have an out-edge.

Consider a two node queueing network, each node with 2 or fewer parallel servers.

If both nodes are involved in the deadlock, so there is a customer in node 1 blocked from entering node 2, and a customer from node 2 blocked from entering node 1, then all servers in node 1 and node 2 in  $D(t)$  will have out edges as they are occupied by a blocked individual. The servers of node 1 and 2 consist of the entirety of  $D(t)$ , and so there is no sink nodes.

Now consider the case when only one node is involved in the deadlock. Without loss of generality, let's say that node 1 is in deadlock with itself, then the servers of node 1 have out-edges. For the servers of node 2 to be part of that weakly connected component, there either needs to be an edge from a server in node 1 to a server in node 2, or an edge from a server in node 2 to a server in node 1. An edge from a server in node 1 to a server in node 2 implies that a customer from node 1 is blocked from entering node 2, and so node 1 is not in deadlock with itself. An edge from a server in node 2 to a server in node 1 implies that a customer in node 2 is blocked from entering node 1. In this case one server in node 2 has an out-edge. Now either the other server of node two is still in server, and so isn't part of that weakly connected component, or the other server's customer is blocked and so has an out edge.

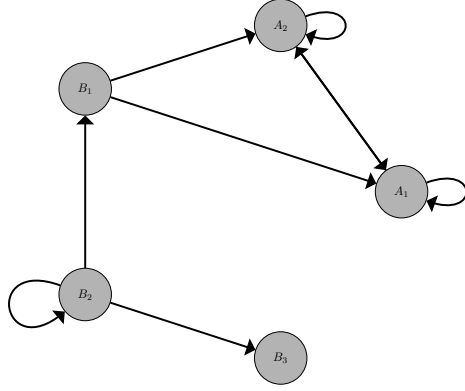
For the case of a queueing network with two nodes and more than 2 servers at node 2, consider the following counter-example:

Node  $A$  has two parallel server, node  $B$  has three parallel servers. Begin with all servers occupied by customers in service and full queues. The customer at server  $A_1$  is blocked to node  $A$ . The customer at server  $B_1$  is blocked to node  $A$ . The customer at server  $B_2$  is blocked to node  $B$ . The customer at server  $A_2$  is blocked to node  $A$ . The resulting state digraph in Figure 4a has a weakly connected component with a sink.

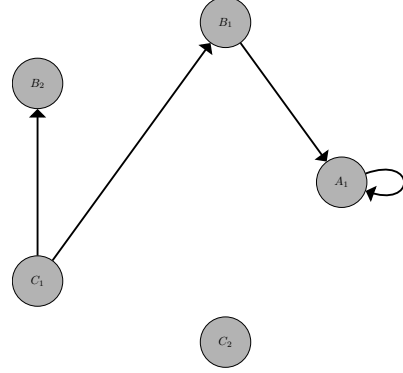
Consider a queueing network with  $N$  nodes, each with a single server.

If  $1 \leq n \leq N$  nodes are involved in the deadlock, then each server in those  $n$  nodes has a blocked customer, and so has an out-edge. Of the other nodes, they can only be in the same weakly connected component if either they contain a individual blocked by those in deadlock, in which case they will have an out edge; or they contain a blocked individual blocked by those directly or indirectly blocked by those in deadlock, in which case they will have an out edge; or they are blocking someone who is blocked directly or indirectly by those in deadlock. However this last case cannot happen, as every node is single server each person can only be blocked by one other individual at a time.

For the case of a queueing network with more than two nodes with multiple servers, the following counter-example proves the claim:



(a) State Digraph of Counter-Example 1.



(b) State Digraph of Counter-Example 2.

Figure 4: Counter examples.

Node  $A$  has one parallel server, node  $B$  has two parallel servers, and node  $C$  has three parallel servers. Begin with all servers occupied by customers in service and full queues. The customer at server  $B_1$  is blocked from entering node  $A$ . Then the customer at server  $C_1$  is blocked from entering node  $B$ . Then the customer at server  $A_1$  is blocked from entering node  $A$ . The resulting state digraph in Figure 4b has a weakly connected component with a sink.

□

## 4 Simulation Model

A simulation model is used to run a discrete event simulation of the open restricted queueing network, detect deadlock and stop the simulation at this point. The time the simulation ran for until deadlock is recorded. The model is generic, and can handle any number of service center nodes, each defined by its external arrival rate, service rate, number of servers and queueing capacity, which may be set to  $\infty$  is needed. A routing matrix with entries  $r_{ij}$  describes how every node is connected to every other node.

The simulation model is built in an object-orientated manor using Python. Customers, nodes, servers and the queueing network itself are objects that interact or contain each other.

The state digraph  $D(t)$  described in Section 3 is an attribute of the queueing network. Edges are drawn or removed whenever the following events take place: a customer begins service at a server, a customer becomes blocked, a customer leaves a server. The model is a discrete event simulation, and the deadlocked condition is checked immediately before the clock is moved on. Here a brute force algorithm is used to check whether each strongly connected component of  $D(t)$  is a knot. Once a knot is found the simulation terminates and the time elapsed recorded.

## 5 Markovian Models of Deadlocking Queueing Networks

In this section markov models are build for the following deadlocking queueing networks

- Open one node, single server restricted queueing network with feedback loop.
- Open two node, single server restricted queueing network with routes between nodes.
- Open two node, single server restricted queueing network with routes between nodes and feedback loops.
- Open one node, multi server restricted queueing network with feedback loop.
- Open two node, mutli server restricted queueing network with routes between nodes.

In general a continuous Markov chain model of a deadlocking queueing network is a set of states  $s \in S$  and the transition rates between these states  $q_{s_1, s_2}$ . Each state  $s$  uniquely defines a configuration of customers around the queueing network. Deadlocked states are also present, either denoted by that specific configuration of customers, or by negative numbers, for example  $-1$ . Any deadlocked states cannot transition to any other state, and so is an absorbing state of the Markov chain. Therefore any queueing network that can experience deadlock is guaranteed to experience deadlock, as absorbing Markov chains are guaranteed to enter one of its absorbing states.

The expected time until deadlock is reached is equivalent to the expected time to absorbtion of the Markov chain, which can easily be found [15]. The canonical form of an absorbing Markov chain is

$$P = \begin{pmatrix} T & U \\ 0 & I \end{pmatrix}$$

where  $I$  is an identity matrix.

Now the expected number of time steps until absorption starting from state  $i$  is the  $i$ th element of the vector

$$(I - T)^{-1}e \tag{1}$$

where  $e$  is a vector of 1s.

Therefore by discretising the continuous Markov chain and ensuring the correct order of states, the number expected of time steps to absorption, or deadlock can be found.

When there are more than one deadlock state, there is more than one absorbing state in the Markov chain. The probabilities of which absorbing state a Markov chain will reach can be found. Now the  $(i, j)^{\text{th}}$  element of the vector

$$(I - T)^{-1}U \tag{2}$$

corresponds to the probability of reaching absorbing state  $j$  from transient state  $i$ .



## 5.1 One Node Single Server

Consider the open one node single server restricted queueing network with feedback loop shown in Figure 5. This shows an  $M/M/1$  queue with room for  $n$  customers to queue at any one time, customers arrive at a rate of  $\Lambda$  and served at a rate  $\mu$ . Once a customer has finished service they rejoin the queue with probability  $r_{11}$ , and so exit the system with probability  $1 - r_{11}$ .

Let this system be denoted by  $\Omega_1$  with parameter set  $(\Lambda, \mu, n, r_{11})$ , and the time to deadlock of this system be denoted by  $\omega_1$ .

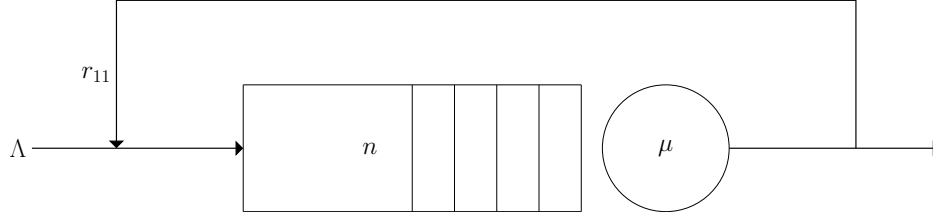


Figure 5: An open one node single server restricted queueing network.

State space:

$$S = \{i \in \mathbb{N} \mid 0 \leq i \leq n + 1\} \cup \{-1\}$$

where  $i$  denotes the number of individuals in service or waiting, and  $-1$  denotes the deadlocked state.

If we define  $\delta = i_2 - i_1$  for all  $i_k \geq 0$  then the transitions are given by:

$$q_{i_1, i_2} = \begin{cases} \Lambda & \text{if } i < n + 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } \delta = 1$$

$$\begin{cases} (1 - r_{11})\mu & \text{if } \delta = -1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } \delta = -1$$

$$q_{i_1, i_2} = \begin{cases} \Lambda & \text{if } i < n + 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } \delta = 1$$

$$\begin{cases} (1 - r_{11})\mu & \text{if } \delta = -1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } \delta = -1$$
(3)

$$q_{i, -1} = \begin{cases} r_{11}\mu & \text{if } i = n + 1 \\ 0 & \text{otherwise} \end{cases} \quad (4)$$

and

$$q_{-1, i} = 0 \quad (5)$$

The Markov chain is shown in Figure 6.

Figure 7 shows the effect of varying the parameters of the queueing network on times to deadlock. Base parameters of  $\Lambda = 10$ ,  $n = 3$ ,  $\mu = 5$  and  $r_{11} = 0.25$  were used.

It can be seen that increasing the arrival rate  $\Lambda$  and the transition probability  $r_{11}$  results in reaching deadlock faster. This is intuitive as increasing these parameters results in the first node's queue filling up quicker.

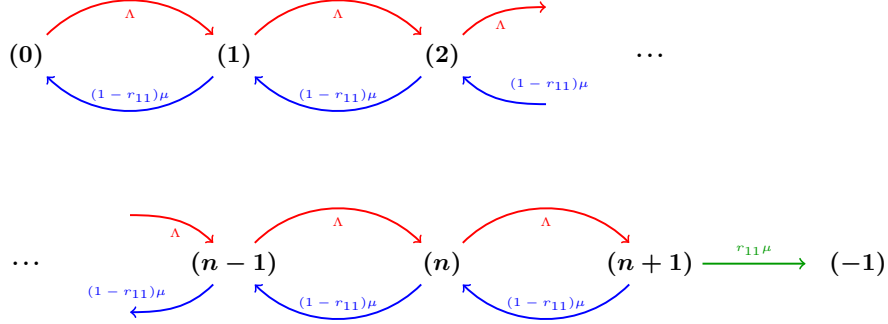


Figure 6: Diagrammatic representation of the Markov chain for  $\Omega_1$ .

Increasing the queueing capacity  $n$  results in reaching deadlock slower. Again this is intuitive, as increasing the queueing capacity allows more customers in the system before becoming deadlock.

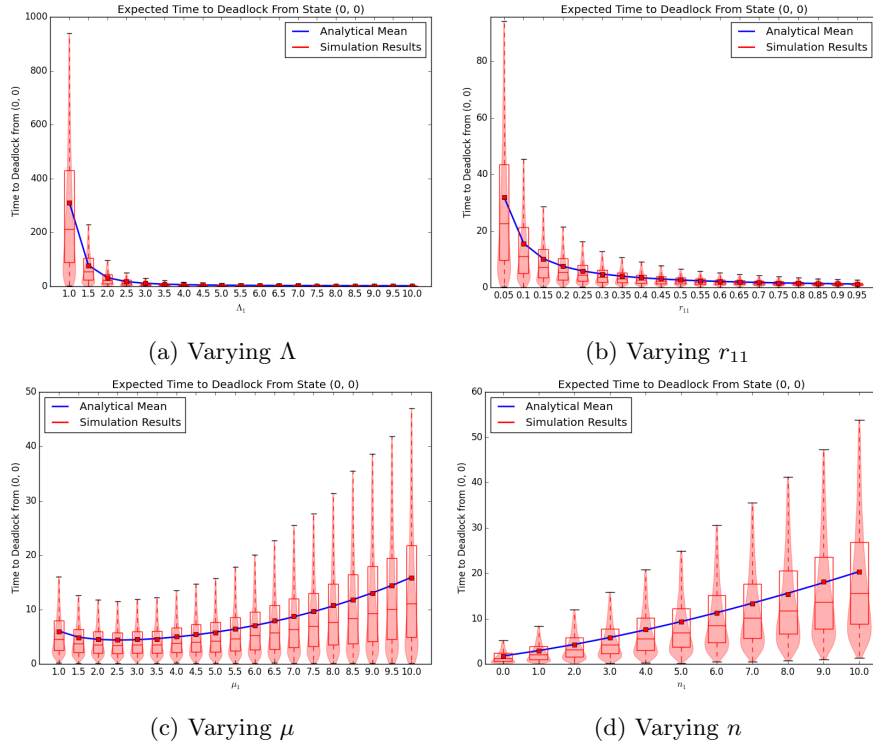


Figure 7: Time to deadlock in  $\Omega_1$ , analytical & simulation results (10,000 iterations).

The behaviour as the service rate  $\mu$  varies is not monotonic, as the service rate contributed towards both moving customers from the system and allowing customers to rejoin the queue, causing blockages and deadlock. This behaviour is described in the following lemma.

**Lemma 1.** *The function  $\omega_1(\mu)$  that describes the time to deadlock of an  $\Omega_1$  system as the service rate  $\mu$  varies, and all other parameters are fixed, has one critical point and is a local minimum for  $\mu \in (0, \infty)$ .*

*Proof.* This proof is by interpretation. The behaviour of  $\omega_1(\mu)$  can be interpreted as follows:

- At  $\lim_{\mu \rightarrow 0} \omega_1(\mu)$  there is infinite service time, and so infinite time until deadlock.
- At  $\lim_{\mu \rightarrow \infty} \omega_1(\mu)$  there is zero service time, the queue can never fill up, and so infinite time to deadlock.
- At low service rates below a certain threshold  $\hat{\mu}$ , the arrival rate is relatively large compared to the service rate, and we can assume a saturated system. At this point services where a customer exits the system does not have much of an effect, as we can assume another arrival immediately. However services where a customer wishes to rejoin the queue results in a blockage as the system is saturated. Therefore, increasing the service rate here increases the chance of a blockage, and so the chance of deadlock.
- Above  $\hat{\mu}$  the service rate is large enough that we cannot assume a saturated system, and so services where the customer exits the system does have an affect on the number of customers in the system. Thus increasing the service rate removes people from the system, and as such there is less chance of getting blocked and deadlocked.

□

**Remark 1.**  $\arg \max_{\mu \in [a,b]} \omega_1(\mu)$  is either  $a$  or  $b$ .

From the closed interval method of finding absolute maximum [17], the absolute maximum of  $\omega_1(\mu)$  on the closed interval  $[a, b]$  is either the critical points in  $(a, b)$ ,  $a$  or  $b$ . The critical points in  $(a, b)$  is  $\hat{\mu}$ , and is a local minimum (from Lemma 1), and so  $\hat{\mu} \leq a$  and  $\hat{\mu} \leq b$ . Therefore  $\omega_1(\mu)$  obtains its maximum at either  $a$  or  $b$ .

## 5.2 Two Node Single Server without Self-Loops

Consider the open two node single server restricted queueing network shown in Figure 8. This shows two  $M/M/1$  queues, with  $n_i$  queueing capacity at each service station and service rates  $\mu_i$ .  $\Lambda_i$  is the external arrival rates to each service station. All routing possibilities except self loops are possible, where the routing probability from node  $i$  to node  $j$  is denoted by  $r_{ij}$ .

Let this system be denoted by  $\Omega_2$  with parameter set  $(\Lambda_1, \Lambda_2, \mu_1, \mu_2, n_1, n_2, r_{12}, r_{21})$ .

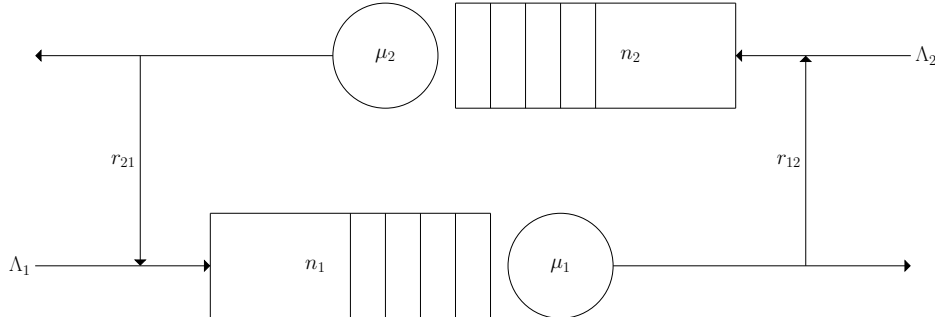


Figure 8: An open two node single server restricted queueing network.

- State space:

$$S = \{(i, j) \in \mathbb{N}^{(n_1+2 \times n_2+2)} \mid 0 \leq i + j \leq n_1 + n_2 + 2\} \cup \{(-1)\}$$

where  $i$  denotes the number of individuals:

- In service or waiting at the first node.
- Occupying a server but having finished service at the second node waiting to join the first.

where  $j$  denotes the number of individuals:

- In service or waiting at the second node.
- Occupying a server but having finished service at the first node waiting to join the second.

and the state  $(-1)$  denotes the deadlocked state.

If we define  $\delta = (i_2, j_2) - (i_1, j_1)$  for all  $(i_k, j_k) \in S$ , then the transitions are given by:

$$q_{(i_1, j_1), (i_2, j_2)} = \left\{ \begin{array}{ll} \left. \begin{array}{l} \Lambda_1 \quad \text{if } i_1 < n_1 + 1 \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (1, 0) \\ \left. \begin{array}{l} \Lambda_2 \quad \text{if } j_1 < n_2 + 1 \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (0, 1) \\ \left. \begin{array}{l} (1 - r_{12})\mu_1 \quad \text{if } j_1 < n_2 + 2 \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (-1, 0) \\ \left. \begin{array}{l} (1 - r_{21})\mu_2 \quad \text{if } i_1 < n_1 + 2 \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (0, -1) \\ \left. \begin{array}{l} r_{12}\mu_1 \quad \text{if } j_1 < n_2 + 2 \text{ and } (i_1, j_1) \neq (n_1 + 2, n_2) \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (-1, 1) \\ \left. \begin{array}{l} r_{21}\mu_2 \quad \text{if } i_1 < n_1 + 2 \text{ and } (i_1, j_1) \neq (n_1, n_2 + 2) \\ 0 \quad \text{otherwise} \end{array} \right\} & \text{if } \delta = (1, -1) \\ 0 & \text{otherwise} \end{array} \right. \quad (6)$$

$$q_{(i_1, j_1), (-1)} = \left\{ \begin{array}{ll} r_{21}\mu_2 & \text{if } (i, j) = (n_1, n_2 + 2) \\ r_{12}\mu_1 & \text{if } (i, j) = (n_1 + 2, n_2) \\ 0 & \text{otherwise} \end{array} \right. \quad (7)$$

and

$$q_{-1, s} = 0 \quad (8)$$

For  $n_1 = 1$  and  $n_2 = 2$ , the resulting Markov chain is shown in Figure 9.

Figure 10 shows the effect of varying the parameters of the above Markov model. Base parameters of  $\Lambda_1 = 4$ ,  $\Lambda_2 = 5$ ,  $n_1 = 3$ ,  $n_2 = 2$ ,  $\mu_1 = 10$ ,  $\mu_2 = 8$ ,  $r_{12} = 0.25$  and  $r_{21} = 0.15$  were used. similar behaviour to Figure 7 can be seen.

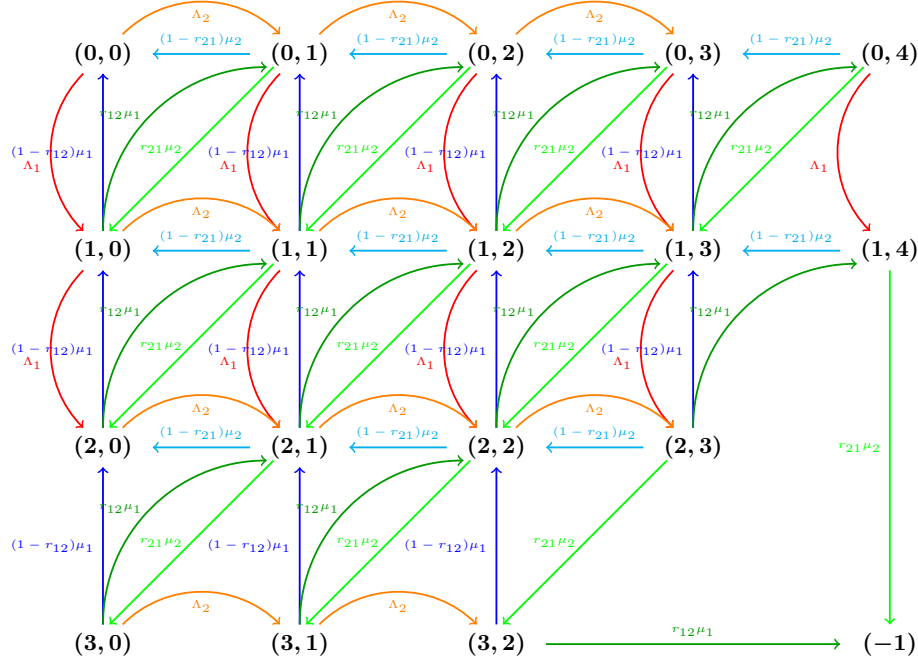


Figure 9: Diagrammatic representation of the Markov chain for  $\Omega_2$  with  $n_1 = 1$  and  $n_2 = 2$ .

### 5.3 Two Node Single Server with Self-Loops

Consider the open two node single server restricted queueing network shown in Figure 11. This shows two  $M/M/1$  queues, with  $n_i$  queueing capacity at each service station and service rates  $\mu_i$ .  $\Lambda_i$  is the external arrival rates to each service station. All routing possibilities are possible, where the routing probability from node  $i$  to node  $j$  is denoted by  $r_{ij}$ .

Let this system be denoted by  $\Omega$  with parameter set  $(\Lambda_1, \Lambda_2, \mu_1, \mu_2, n_1, n_2, r_{11}, r_{12}, r_{21}, r_{22})$ .

- State space:

$$S = \{(i, j) \in \mathbb{N}^{(n_1+2 \times n_2+2)} \mid 0 \leq i + j \leq n_1 + n_2 + 2\} \cup \{(-1), (-2), (-3)\}$$

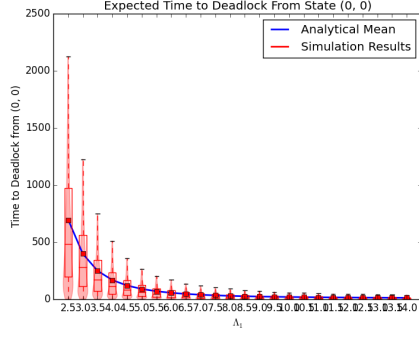
where  $i$  denotes the number of individuals:

- In service or waiting at the first node.
- Occupying a server but having finished service at the second node waiting to join the first.

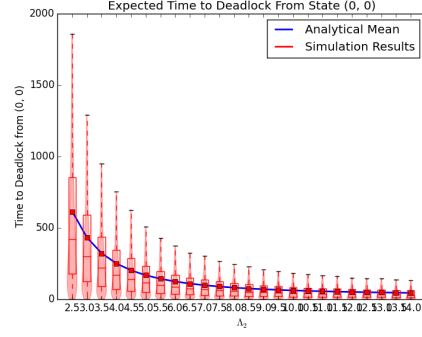
where  $j$  denotes the number of individuals:

- In service or waiting at the second node.
- Occupying a server but having finished service at the first node waiting to join the second.

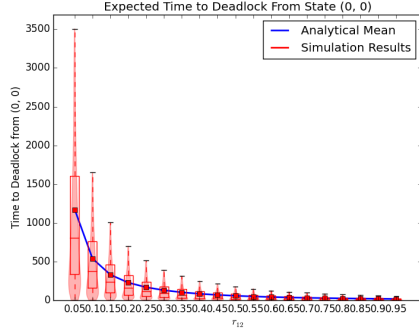
and the state  $(-3)$  denotes the deadlocked state cause by both nodes;  $(-1)$  denotes the deadlocked



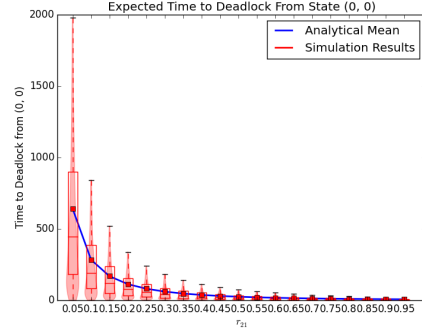
(a) Varying  $\Lambda_1$



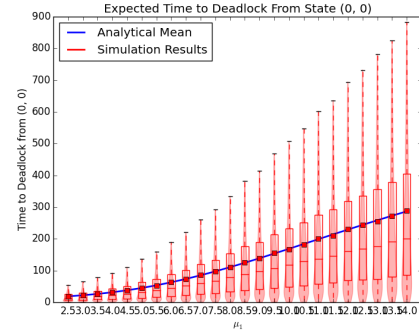
(b) Varying  $\Lambda_2$



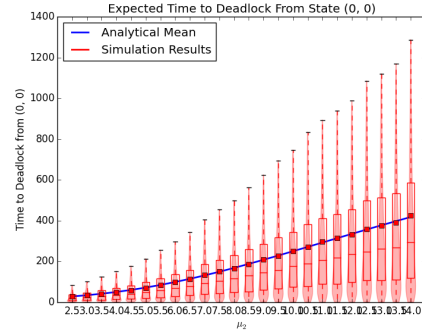
(c) Varying  $r_{12}$



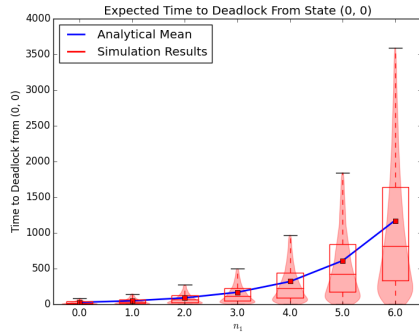
(d) Varying  $r_{21}$



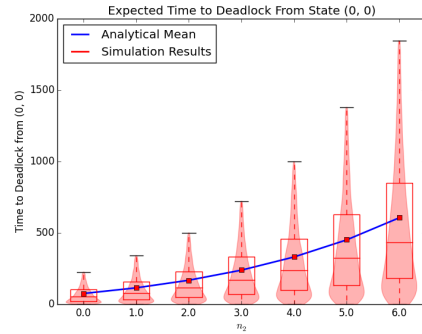
(e) Varying  $\mu_1$



(f) Varying  $\mu_2$



(g) Varying  $n_1$



(h) Varying  $n_2$

Figure 10: Time to deadlock in  $\Omega_2$ , analytical & simulation results (10,000 iterations).

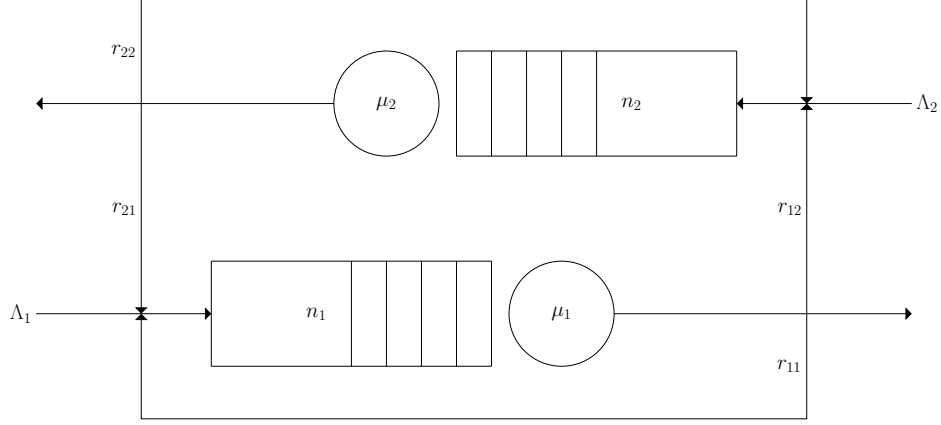


Figure 11: An open two node single server restricted queueing network.

state caused by the first node only; and  $(-2)$  denotes the deadlocked state caused by the second node only.

If we define  $\delta = (i_2, j_2) - (i_1, j_1)$  for all  $(i_k, j_k) \in S$ , then the transitions are given by:

$$q_{(i_1, j_1), (i_2, j_2)} = \left\{ \begin{array}{ll} \Lambda_1 & \text{if } i_1 < n_1 + 1 \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (1, 0)$$

$$\left\{ \begin{array}{ll} \Lambda_2 & \text{if } j_1 < n_2 + 1 \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (0, 1)$$

$$\left\{ \begin{array}{ll} (1 - r_{11} - r_{12})\mu_1 & \text{if } j_1 < n_2 + 2 \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (-1, 0)$$

$$\left\{ \begin{array}{ll} (1 - r_{21} - r_{22})\mu_2 & \text{if } i_1 < n_1 + 2 \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (0, -1)$$

$$\left\{ \begin{array}{ll} r_{12}\mu_1 & \text{if } j_1 < n_2 + 2 \text{ and } (i_1, j_1) \neq (n_1 + 2, n_2) \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (-1, 1)$$

$$\left\{ \begin{array}{ll} r_{21}\mu_2 & \text{if } i_1 < n_1 + 2 \text{ and } (i_1, j_1) \neq (n_1, n_2 + 2) \\ 0 & \text{otherwise} \end{array} \right\} \quad \text{if } \delta = (1, -1)$$

$$0 \quad \text{otherwise}$$
(9)

$$q_{(i_1, j_1), (-1)} = \left\{ \begin{array}{ll} r_{11}\mu_1 & \text{if } i > n_1 \text{ and } j < n_2 + 2 \\ 0 & \text{otherwise} \end{array} \right. \quad (10)$$

$$q_{(i_1, j_1), (-2)} = \left\{ \begin{array}{ll} r_{22}\mu_2 & \text{if } j > n_2 \text{ and } i < n_1 + 2 \\ 0 & \text{otherwise} \end{array} \right. \quad (11)$$

$$q_{(i_1, j_1), (-3)} = \left\{ \begin{array}{ll} r_{21}\mu_2 & \text{if } (i, j) = (n_1, n_2 + 2) \\ r_{12}\mu_1 & \text{if } (i, j) = (n_1 + 2, n_2) \\ 0 & \text{otherwise} \end{array} \right. \quad (12)$$

and

$$q_{-1,s} = 0 \quad (13)$$

$$q_{-2,s} = 0 \quad (14)$$

$$q_{-3,s} = 0 \quad (15)$$

Note that there are only two differences between this formulation and the formulation given in Subsection 5.2: the probabilities of leaving nodes 1 and 2 are now  $(1 - r_{11} - r_{12})\mu_1$  and  $(1 - r_{21} - r_{22})\mu_2$ ; and there are now two more ways to reach deadlock, Equation 10 and Equation 11.

For  $n_1 = 1$  and  $n_2 = 2$ , the resulting Markov chain is shown in Figure 12.

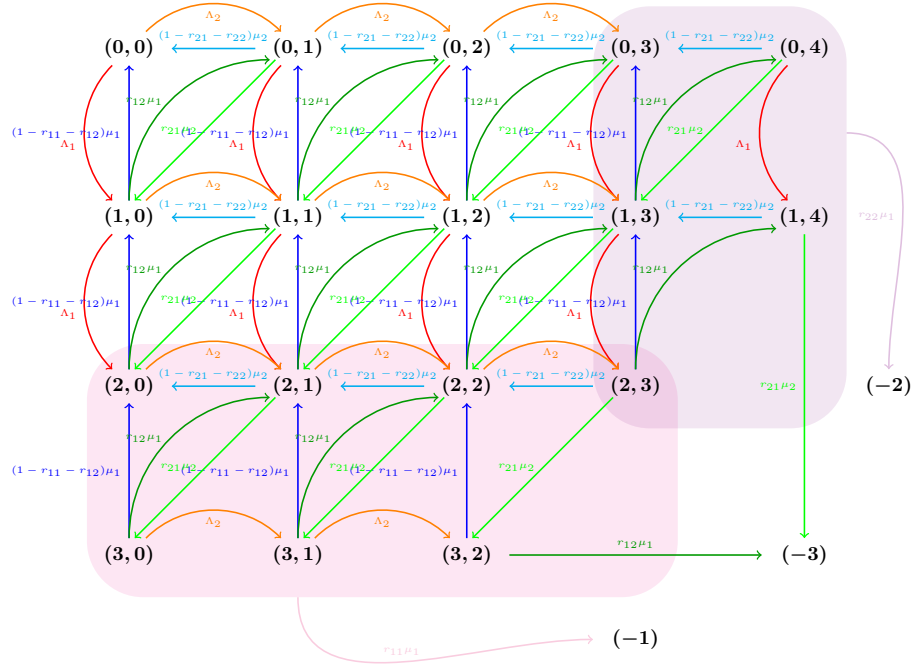


Figure 12: Diagrammatic representation of the Markov chain for  $\Omega$  with  $n_1 = 1$  and  $n_2 = 2$ .

Figure 13 shows the effect of varying the parameters of the above Markov model. Base parameters of  $\Lambda_1 = 4$ ,  $\Lambda_2 = 5$ ,  $n_1 = 3$ ,  $n_2 = 2$ ,  $\mu_1 = 10$ ,  $\mu_2 = 8$ ,  $r_{11} = 0.1$ ,  $r_{12} = 0.25$ ,  $r_{21} = 0.15$  and  $r_{22} = 0.1$  were used.

## 5.4 One Node Multi Server

Consider the open one node multi server restricted queueing network shown in Figure 14. This is an  $\Omega_1$  system with  $c$  parallel servers.



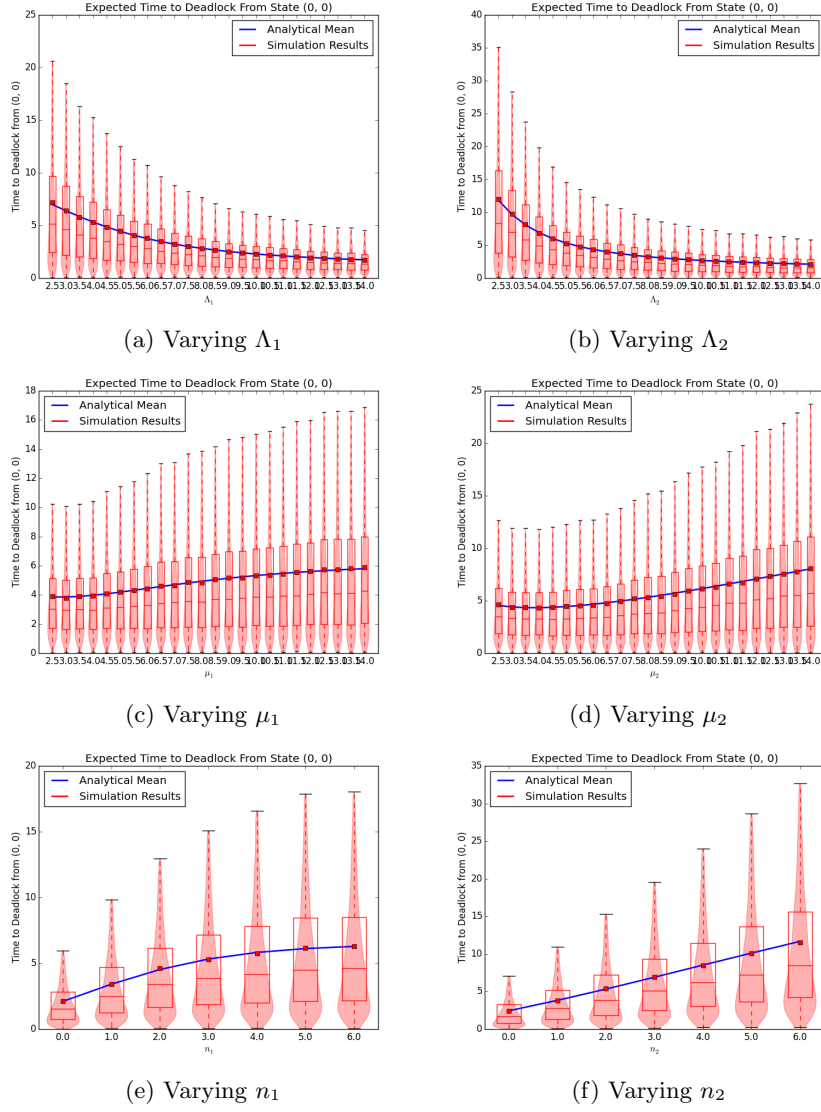


Figure 13: Time to deadlock in  $\Omega$ , analytical & simulation results (10,000 iterations).

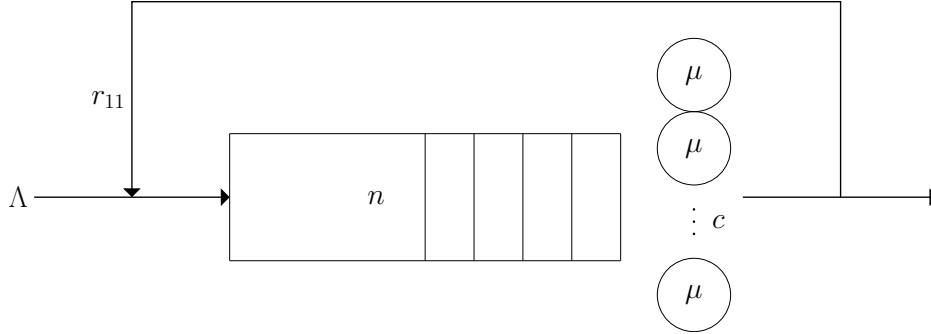


Figure 14: An open one node multi server restricted queueing network.

State space:

$$S = \{i \in \mathbb{N} \mid 0 \leq i \leq n + 2c\}$$

where  $i$  denotes the number of individuals at the node plus the number of individuals blocked at that node. For example,  $i = n + c + 2$  denotes a full system,  $n + c$  individuals in the node, and 2 of those individuals are also blocked. The state  $i = n + 2c$  denotes the deadlocked state.

If we define  $\delta = i_2 - i_1$  for all  $i_k \in S$ , then the transitions are given by:

$$q_{i_1, i_2} = \begin{cases} \Lambda & \text{if } \delta = 1 \\ (1 - r_{11})\mu \min(i, c) & \text{if } \delta = -1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } i_1 < n + c \quad (16)$$

$$q_{i_1, i_2} = \begin{cases} (c - b)r_{11}\mu & \text{if } \delta = 1 \\ (1 - r_{11})(b - k)\mu & \text{if } \delta = -b - 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{if } i_1 = n + c + b \quad \forall \quad 0 \leq b \leq c \quad (17)$$

where  $b$  denotes the number of blocked customers. The Markov chain is shown in Figure 15.

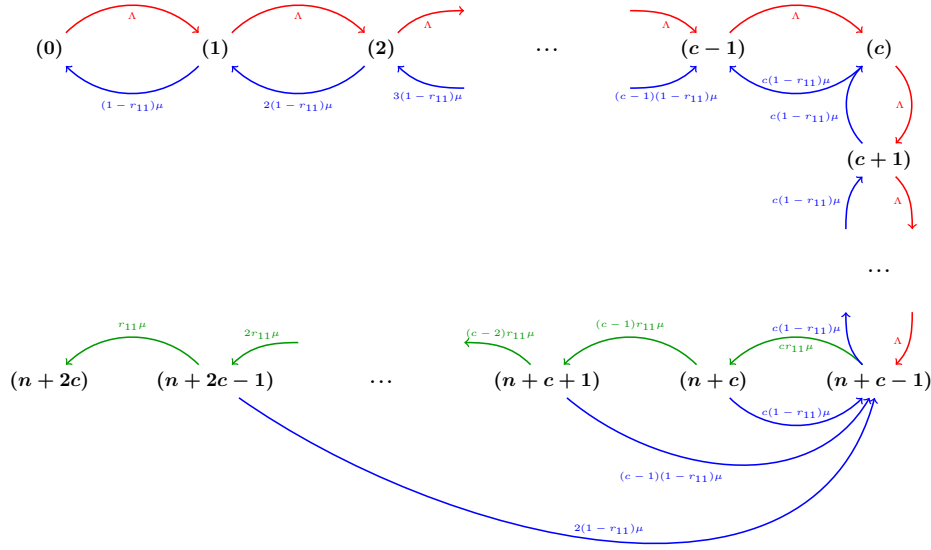


Figure 15: Diagrammatic representation of the Markov chain for a multi server  $\Omega_1$  system.

Increasing the amount of servers has a similar effect to increasing the queueing capacity, there are now more transient states to go through before reaching the deadlocked state. Varying the amount of servers has a greater effect on the time to deadlock however, as any states in which customers are blocked ( $i = n + c + 1$  to  $i = n + 2c$ ) can jump back to state  $i = n + c - 1$  simply with a service and an exit. Increasing the amount of servers also increases the rate at which  $i$  are reduced for most states, but not the rates at which

$i$  is increased.

Figure 16 shows the effect of varying the parameters of the above Markov model. Base parameters of  $\Lambda = 6$ ,  $n = 3$ ,  $\mu = 2$ ,  $r_{11} = 0.5$  and  $c = 2$  were used.

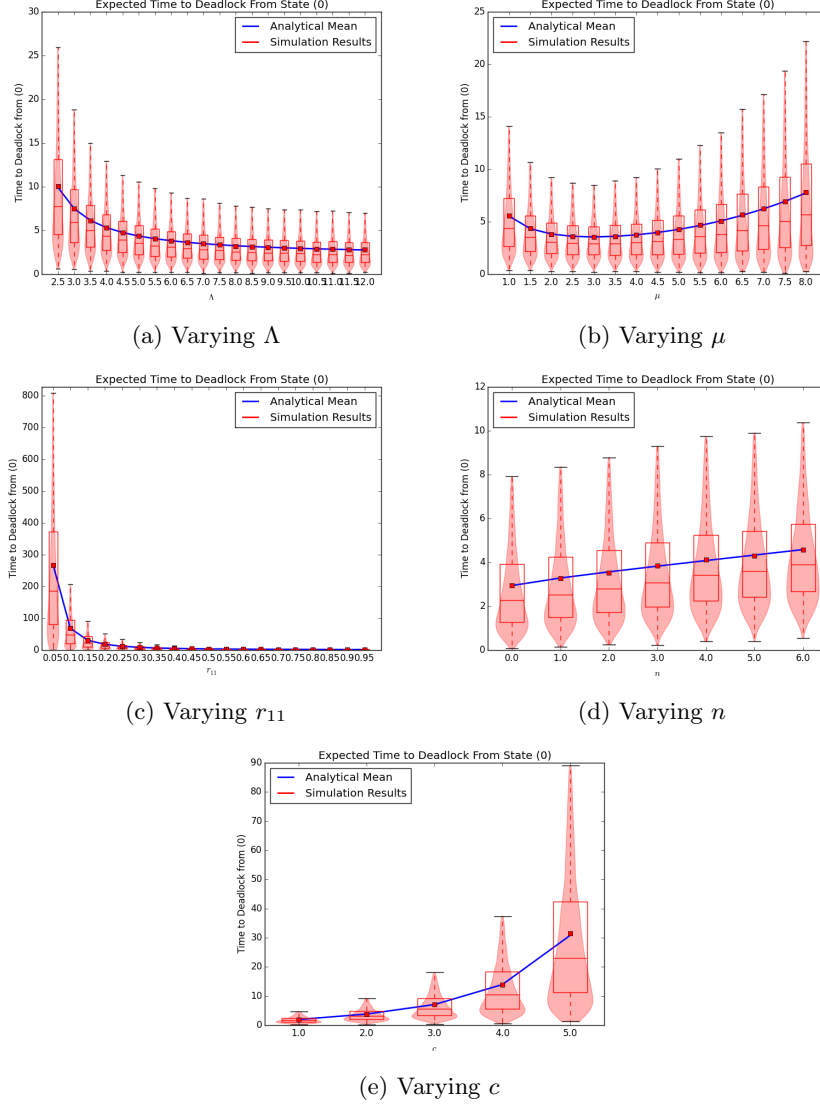


Figure 16: Time to deadlock in multi-server  $\Omega_1$ , analytical & simulation results (10,000 iterations).

## 5.5 Two Node Multi Server without Self-Loops

Consider the open two node multi server restricted queueing network shown in Figure 17. This is an  $\Omega_2$  system with  $c_i$  parallel servers at node  $i$ .

State space:

$$S = \{(i, j) \in \mathbb{N}^{(n_1+c_1+c_2) \times (n_2+c_2+c_1)} \mid i \leq n_1 + c_1 + j, j \leq n_2 + c_2 + i\}$$

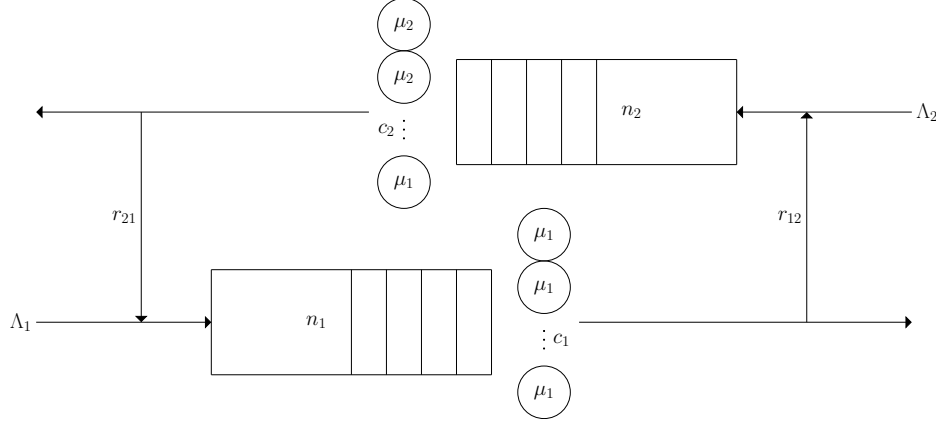


Figure 17: An open two node multi server restricted queueing network.

|                   | $j_1 < n_2 + c_2$  | $j_1 = n_2 + c_2$   | $j_1 > n_2 + c_2$  |
|-------------------|--|---|--|
| $i_1 < n_1 + c_1$ | $\Lambda_1$ if $\delta = (1, 0)$<br>$\Lambda_2$ if $\delta = (0, 1)$<br>$r_{12}s_1\mu_1$ if $\delta = (-1, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, -1)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$ | $\Lambda_1$ if $\delta = (1, 0)$<br>$r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, -1)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$ | $\Lambda_1$ if $\delta = (1, 0)$<br>$r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (0, -1)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (-1, -1)$   |
| $i_1 = n_1 + c_1$ | $\Lambda_2$ if $\delta = (0, 1)$<br>$r_{12}s_1\mu_1$ if $\delta = (-1, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$                                      | $r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$                                      | $r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, 0)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (-1, -1)$  |
| $i_1 > n_1 + c_1$ | $\Lambda_2$ if $\delta = (0, 1)$<br>$r_{12}s_1\mu_1$ if $\delta = (-1, 0)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, -1)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$                                     | $r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-1, -1)$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (0, -1)$                                     | $r_{12}s_1\mu_1$ if $\delta = (0, 1)$<br>$r_{21}s_2\mu_2$ if $\delta = (1, 0)$<br>$(1 - r_{12})s_1\mu_1$ if $\delta = (-\min(b_1 + 1, b_2 + 1), -\min(b_1, b_2 + 1))$<br>$(1 - r_{21})s_2\mu_2$ if $\delta = (-\min(b_1 + 1, b_2), -\min(b_1 + 1, b_2 + 1))$ |

Table 1: Table of transitions for a multi server two node network.

where  $i$  denotes the number of individuals at Node 1 plus the number of individuals blocked waiting to enter Node 1, and  $j$  denotes the number of individuals at Node 2 plus the number of individuals blocked waiting to enter Node 2. For example,  $(i, j) = (n_1 + c_1 + 2, n_2 + c_2 + 1)$  denotes a full system,  $n_1 + c_1$  individuals at Node 1, two of whom are blocked waiting to enter Node 2;  $n_2 + c_2$  individuals at Node 2, one of whom is blocked waiting to enter Node 1. The state  $(i, j) = (n_1 + c_1 + c_2, n_2 + c_2 + c_1)$  denotes the deadlocked state.

The Markov chain is shown in Figure 18.

Define  $\delta = (i_2, j_2) - (i_1, j_1)$ ,  $b_1 = \max(0, i_1 - n_1 - c_1)$ ,  $b_2 = \max(0, i_2 - n_2 - c_2)$ ,  $s_1 = \min(i_1, c_1) - b_2$  and  $s_2 = \min(i_2, c_2) - b_1$ , then the transitions  $q_{(i_1, j_1), (i_2, j_2)}$  are given by Table 1.

The values  $b_1$  and  $b_2$  correspond to the number of people blocked to Node 1 and Node 2 respectively. The values  $s_1$  and  $s_2$  correspond to the amount of people currently in service at Node 1 and Node 2 respectively.

Figure 19 shows the effect of varying the parameters of the above Markov model. Base parameters of  $\Lambda_1 = 9$ ,  $\Lambda_2 = 7.5$ ,  $n_1 = 2$ ,  $n_2 = 1$ ,  $\mu_1 = 5.5$ ,  $\mu_2 = 6.5$ ,  $r_{12} = 0.7$ ,  $r_{21} = 0.6$ ,  $c_1 = 2$  and  $c_2 = 2$  were used.

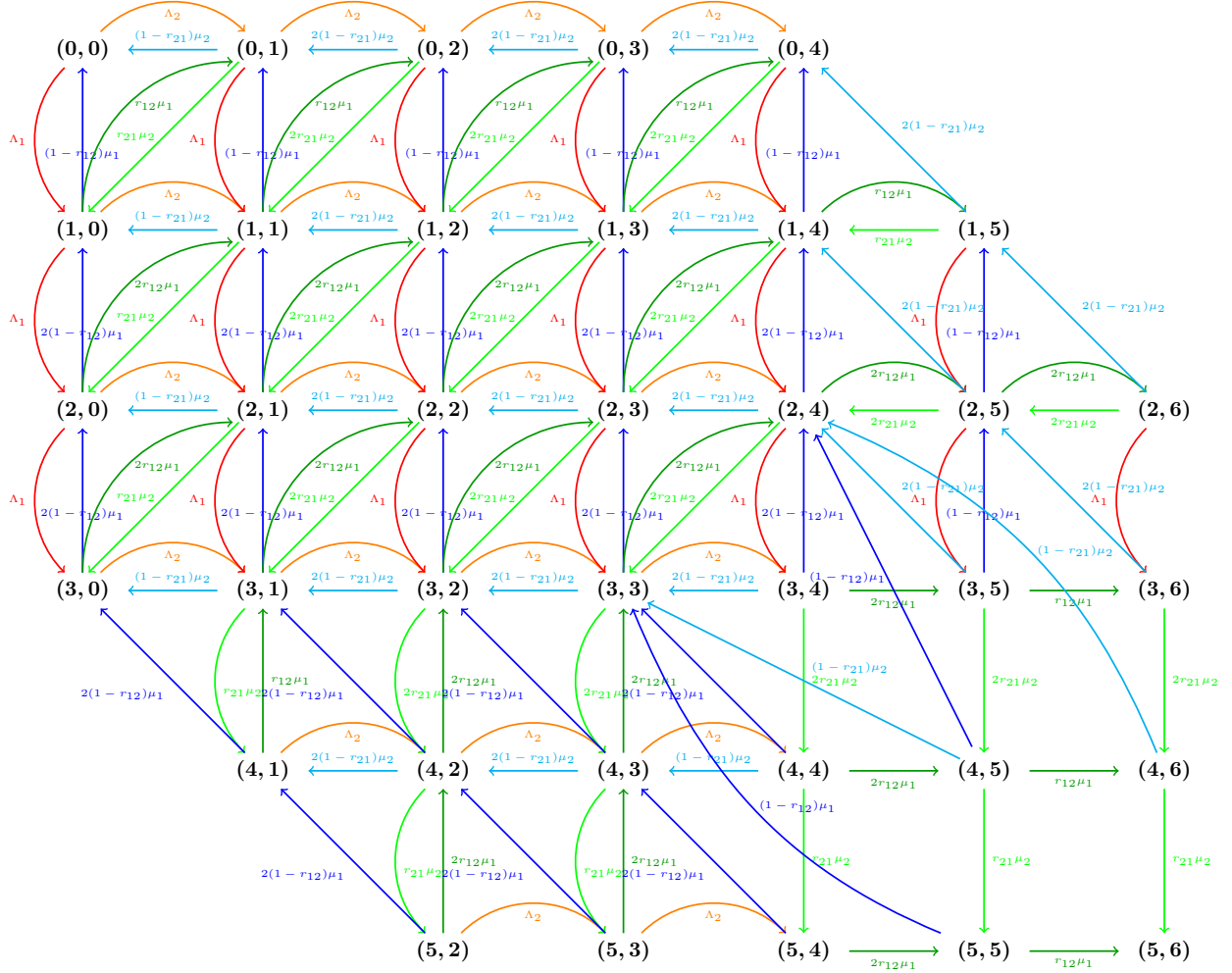


Figure 18: Diagrammatic representation of the Markov chain for a multi server  $\Omega_2$  system with  $n_1 = 1$ ,  $n_2 = c_1 = c_2 = 2$ . The deadlocked state is  $(5, 6)$ .

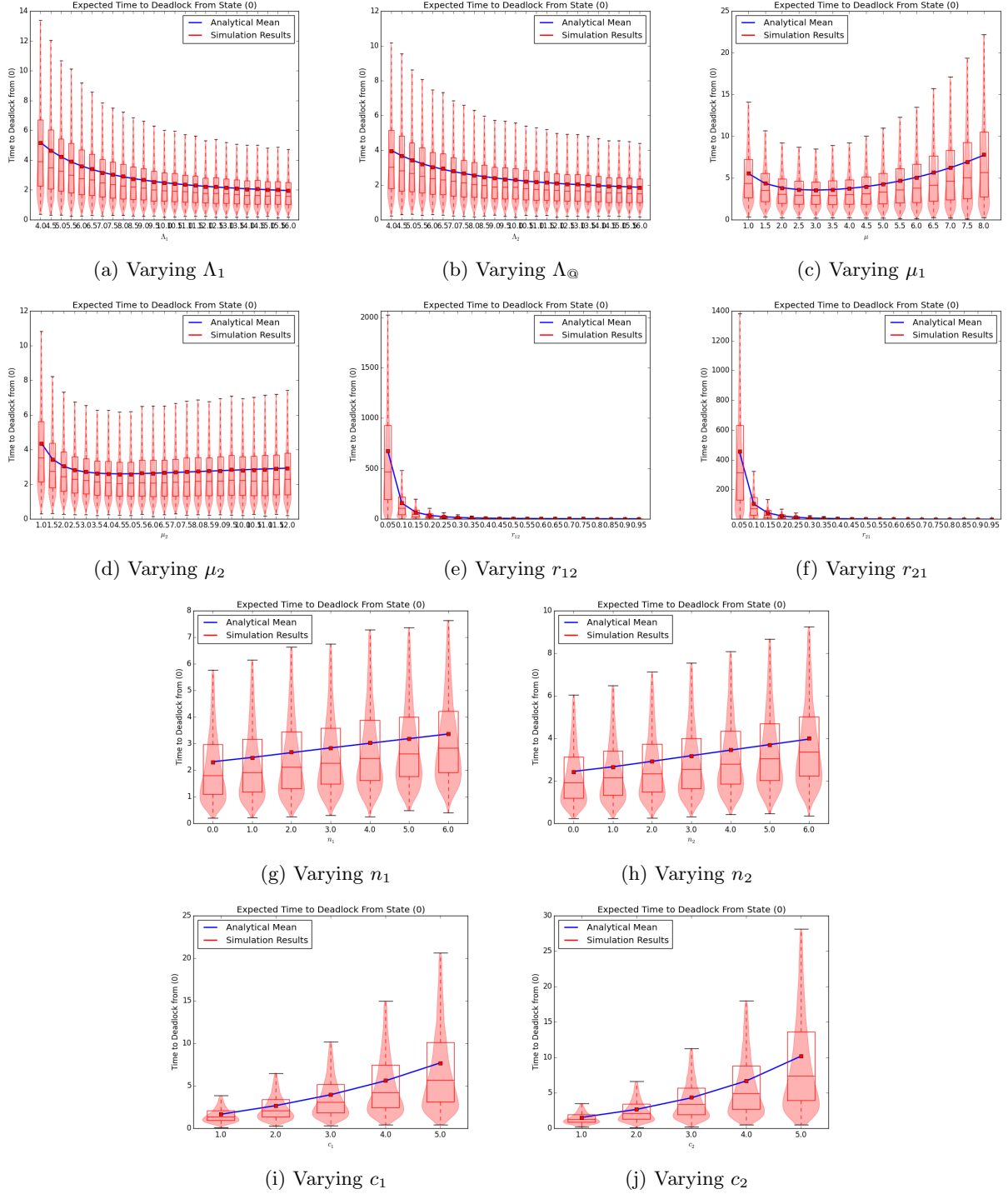


Figure 19: Time to deadlock in multi-server  $\Omega_2$ , analytical & simulation results (10,000 iterations).

## 6 A Bound on the Time to Deadlock

This section gives a bound on the mean time to deadlock of the  $\Omega$  system. In order to do this, six deadlocking queueing networks are defined as follows, all single server:

- Define  $\Omega_{1_1}^*$  as the 1 node queueing network described in Subsection 5.1 with the parameter set  $(\Lambda_1, \mu_1, n_1, r_{11})$ . Let its mean time to deadlock be denoted by  $\omega_{1_1}^*$ .
- Define  $\Omega_{1_1}^{**}$  as the 1 node queueing network described in Subsection 5.1 with the parameter set  $(\Lambda_1, m_1, n_1, r_{11})$ . Let its mean time to deadlock be denoted by  $\omega_{1_1}^{**}$ .
- Define  $\Omega_{1_2}^*$  as the 1 node queueing network described in Subsection 5.1 with the parameter set  $(\Lambda_2, \mu_2, n_2, r_{22})$ . Let its mean time to deadlock be denoted by  $\omega_{1_2}^*$ .
- Define  $\Omega_{1_2}^{**}$  as the 1 node queueing network described in Subsection 5.1 with the parameter set  $(\Lambda_2, m_2, n_2, r_{22})$ . Let its mean time to deadlock be denoted by  $\omega_{1_2}^{**}$ .
- Define  $\Omega_2$  as the 2 node queueing network described in Subsection 5.2 with the parameter set  $(\Lambda_1, \Lambda_2, \mu_1, \mu_2, n_1, n_2, r_{12}, r_{21})$ . Let its mean time to deadlock be denoted by  $\omega_2$ .
- Define  $\Omega$  as the 2 node queueing network described in Subsection 5.3 with the parameter set  $(\Lambda_1, \Lambda_2, \mu_1, \mu_2, n_1, n_2, r_{11}, r_{12}, r_{21}, r_{22})$ . Let its mean time to deadlock be denoted by  $\omega$ .

where  $m_1 = \frac{\mu_2}{3+2\frac{\mu_2}{\mu_1} + \mu_2}$ , and  $m_2 = \frac{\mu_1}{3+2\frac{\mu_1}{\mu_2} + \mu_1}$ .

Figure 20 shows how  $\Omega$  contains, and is made up by,  $\Omega_{1_1}$ ,  $\Omega_{1_2}$  and  $\Omega_2$ .

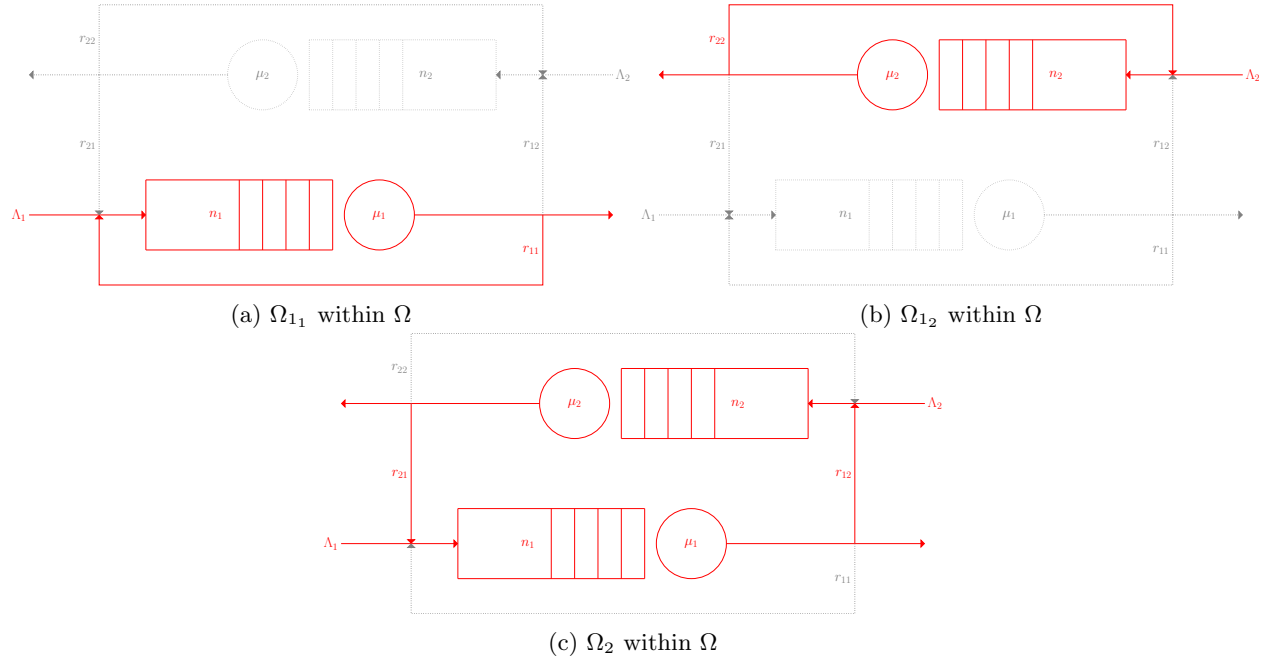


Figure 20: Decomposition of  $\Omega$  into  $\Omega_{1_1}$ ,  $\Omega_{1_2}$  and  $\Omega_2$

Defining  $\omega_{1_j} = \max(\omega_{1_j}^*, \omega_{1_j}^{**})$  we get the following bound:

**Theorem 2.** *For any parameter sets the following inequality holds:  $\omega \leq \min(\omega_{1_1}, \omega_{1_2}, \omega_2)$*

*Proof.* First, define the following systems:

- Let  $\tilde{\Omega}_{1_1}$  denote the  $\Omega_{1_1}$  system embedded within  $\Omega$ . Let  $\tilde{\omega}_{1_1}$  denote the mean time to deadlock of  $\tilde{\Omega}_{1_1}$ .
- Let  $\tilde{\Omega}_{1_2}$  denote the  $\Omega_{1_2}$  system embedded within  $\Omega$ . Let  $\tilde{\omega}_{1_2}$  denote the mean time to deadlock of  $\tilde{\Omega}_{1_2}$ .
- Let  $\tilde{\Omega}_2$  denote the  $\Omega_2$  system embedded within  $\Omega$ . Let  $\tilde{\omega}_2$  denote the mean time to deadlock of  $\tilde{\Omega}_2$ .

Now  $\tilde{\omega}_{1_1}$  is the mean time to state (-1) in  $\Omega$ ,  $\tilde{\omega}_{1_2}$  is the mean time to state (-2) in  $\Omega$  and  $\tilde{\omega}_2$  is the mean time to state (-3) in  $\Omega$ . Therefore  $\omega = \min(\tilde{\omega}_{1_1}, \tilde{\omega}_{1_2}, \tilde{\omega}_2)$ , as the mean time to deadlock in  $\Omega$  is the expected time it takes to reach either (-1), (-2) or (-3), whichever comes first.

Comparing  $\tilde{\Omega}_2$  to  $\Omega_2$ : The effective arrival rate to Node 1 in  $\tilde{\Omega}_2$  is greater than or equal to the effective arrival rate to Node 1 in  $\Omega_2$ . This is due to the extra customers who are rejoining the queue after service. Similarly the effective arrival rate to Node 2 in  $\tilde{\Omega}_2$  is greater than or equal to the effective arrival rate to Node 2 in  $\Omega_2$ . As an increase in the arrival rate causes the mean time to deadlock to decrease, we can conclude  $\tilde{\omega}_2 \leq \omega_2$

Consider  $\tilde{\Omega}_{1_1}$ . Consider the expected effective service time, the time that  $\tilde{\Omega}_{1_1}$ 's state does not change due to services or outside factors.

The shortest expected effective service time is equal to  $\frac{1}{\mu_1}$ , corresponding to when neither Node 1 nor Node 2 are full. Therefore the largest effective service rate is  $\mu_1$ .

The longest service time corresponds to when Node 2 is full, Node 1 finishes service and gets blocked. Here the customer in service at Node 1 spends  $\frac{1}{\mu_1}$  in service, then gets blocked. That customer remains blocked until there is room at Node 2, that is when Node 2 has a service. As all service rates are Markovian, the time spent blocked is  $\frac{1}{\mu_2}$ .

If the customer finishing service at Node 2 transitions to Node 1, then  $\tilde{\Omega}_{1_1}$ 's state hasn't changed, and Node 2 is full again. The next customer at Node 1 now spends  $\frac{1}{\mu_1}$  in service. This process will repeat if Node 1 is blocked before Node 2 releases a customer, and so the process is repeated with probability  $P_{\text{repeat}}$ .

And so the longest effective service time is:

$$\begin{aligned}
 \frac{1}{m_1} &= \frac{2}{\mu_1} + \frac{1}{\mu_2} + P_{\text{repeat}} \left( \frac{2}{\mu_1} + \frac{1}{\mu_2} + P_{\text{repeat}} \left( \frac{2}{\mu_1} + \frac{1}{\mu_2} + P_{\text{repeat}} \left( \dots \right. \right. \right. \\
 &= \left( \frac{2}{\mu_1} + \frac{1}{\mu_2} \right) \times (1 + P_{\text{repeat}} + P_{\text{repeat}}^2 + P_{\text{repeat}}^3 + \dots) \\
 &= \left( \frac{2}{\mu_1} + \frac{1}{\mu_2} \right) \times \left( \frac{1}{1 - P_{\text{repeat}}} \right)
 \end{aligned}$$



If  $S_1$  is the time the customer at Node 1 spends in service, and  $S_2$  is the time the customer at node 2 spends in service, then  $S_1 \sim \text{Exp}(\mu_1)$  and  $S_2 \sim \text{Exp}(\mu_2)$ . Now  $P_{\text{repeat}} = P(S_1 < S_2) = \frac{\mu_1}{\mu_1 + \mu_2}$ .

Therefore the smallest effective service rate is:

$$\begin{aligned}
m_1 &= \frac{1}{\left(\frac{2}{\mu_1} + \frac{1}{\mu_2}\right) \left(\frac{1}{1-P_{\text{repeat}}}\right)} \\
&= \frac{1}{\left(\frac{2}{\mu_1} + \frac{1}{\mu_2}\right) \left(\frac{1}{1-\frac{\mu_1}{\mu_1+\mu_2}}\right)} \\
&= \frac{1}{\left(\frac{2}{\mu_1} + \frac{1}{\mu_2}\right) \left(\frac{\mu_1+\mu_2}{\mu_2}\right)} \\
&= \frac{\mu_2}{\left(\frac{2}{\mu_1} + \frac{1}{\mu_2}\right) (\mu_1 + \mu_2)} \\
&= \frac{\mu_2}{3 + 2\frac{\mu_2}{\mu_1} + \frac{\mu_1}{\mu_2}}
\end{aligned}$$

As time to deadlock does not increase monotonically with service rate (see Figure 7c) we cannot simply compare  $\tilde{\Omega}_{1_1}$  with the system that has the largest of smallest service rate; instead we must compare  $\tilde{\Omega}_{1_1}$  with either the system with the smallest service rate or the largest, whichever yields the longest time to deadlock.

The largest service rate is  $\mu_1$ , corresponding to system  $\Omega_{1_1}^*$  and the smallest is  $m$  corresponding to  $\Omega_{1_1}^{**}$ . And so we compare with whichever system takes longer to reach deadlock. Using Lemma 1 and Remark 1 we compare with  $\omega_{1_1} = \max(\omega_{1_1}^*, \omega_{1_1}^{**})$ .

And so comparing this to  $\tilde{\Omega}_{1_1}$ : the effective arrival rate in  $\tilde{\Omega}_{1_1}$  is greater than or equal to the effective arrival rate in that system; this is due to the extra customers who have transitioned from the Node 2 to Node 1. An increase in the arrival rate causes the mean time to deadlock to decrease.

We can conclude that  $\tilde{\omega}_{1_1} \leq \omega_{1_1}$  at  $\tilde{\omega}_{1_1}$ 's worst case scenrio, and so  $\tilde{\omega}_{1_1} \leq \omega_{1_1}$  in general.

Similarly we can conclude  $\tilde{\omega}_{1_2} \leq \omega_{1_2}$ .

And so

$$\begin{aligned}
\min(\tilde{\omega}_{1_1}, \tilde{\omega}_{1_2}, \tilde{\omega}_2) &\leq \min(\omega_{1_1}, \omega_{1_2}, \omega_2) \\
\omega &\leq \min(\omega_{1_1}, \omega_{1_2}, \omega_2)
\end{aligned}$$

□

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