

Chapter 1

Two Sided DQBD Process and Solutions to the Tail Decay Rate Problem and Their Applications to the Generalized Join Shortest Queue

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Abstract We are concerned with a two sided doubly quasi-birth-and-death process. Under a discrete time setting, this is a two dimensional skip free random walk on the half space whose second component is a nonnegative integer valued while its first component may take positive or negative integers. Our major interest is in the tail decay rate of the stationary distribution of this two sided process as either one of the components goes either to infinity or to minus infinity, provided the stationary distribution exists. The author [1] recently obtained two kinds of decay rates, called weak and exact for the doubly QBD, DQBD for short, in terms of the transition kernel of the DQBD. We extend those results to the two sided DQBD, and apply to the generalized shortest queue. The tail decay rate problem for this queueing model has been only partially answered in the literature. We show that a weak decay rate, that is, the decay rate in the logarithmic sense, is completely specified in terms of the primitive data for the generalized shortest queue. This refines results in Miyazawa [2] and corrects some results in Li, Miyazawa and Zhao [3].

1.1 Introduction

A quasi birth-and-death process, QBD process for short, is a continuous time Markov chain which has a main state, called level, and a background state in such a way that the level is nonnegative integer valued, and its increments are ± 1 at most and controlled by the background state. This model has been well studied when the background state space is finite (see, e.g. [4], [5]).

We are concerned with the case that the background space is infinite. Li, Miyazawa and Zhao [3] recently proposed a double sided QBD process for the generalized join shortest queue with two waiting lines, by extending the level of

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such a QBD process to be integer valued. This queue is a service system with two parallel queues that have three arrival streams, two of which are dedicated to each queue and the other of which chooses the shortest queue with tie breaking. Assume that those arrival streams are independent and subject to Poisson processes, and service times are independently, identically and exponentially distributed at each queue. Then, this queue can be formulated as the QBD process or the two sided QBD process. In particular, the latter model is required when we take the difference of the two queues as level.

It is notable that the transition structure may change in the double sided QBD when the level process goes through zero. This is crucial to formulate the generalized join shortest queue as the double sided QBD. In this chapter, we specialize this double sided QBD in such a way that its background process is birth-and-death. We refer to this process as a two sided doubly quasi birth-and-death process, a two sided DQBD for short. Since those QBD and DQBD can be formulated as discrete time Markov chains, we are only concerned with the discrete time processes throughout the chapter.

We are interested in the asymptotic behaviors of the stationary distributions of the level and background state as their values go to infinity, provided it exists. Due to the special structure of the two sided DQBD, the QBD structure is preserved when the level and background are exchanged. So, we mainly consider the asymptotics for the level. We are concerned with two types of the asymptotic decays of the stationary probabilities as the level goes to infinity.

One type is called a weak decay, which is meant that the logarithm of the stationary probability divided by the level n converges to a constant, say $-a$, as n goes to infinity. Then, e^{-a} is simply referred to as a weak decay rate. Another type is called an exactly geometric decay, which is meant that the stationary probability multiplied by a power constant to the level n , say α^n , converges to another constant as n goes to infinity. Then, α^{-1} is referred to as an exactly geometric decay rate. In [1], more general types of exact decay rates are considered, but we are only concerned with these two types of decay rates in this chapter.

The purpose of this chapter is twofold. We first study the decay rate problem for the two sided DQBD process, by extending the approach for the DQBD process in [1]. We completely characterize the weak decay rates in terms of the transition probabilities (Theorems 1.3 and 1.4). For the exactly geometric decay, we find sufficient conditions, which are close to necessary conditions (Theorem 1.3). We secondly apply these results to find the decay rates of the stationary distributions of the minimum of the two queues and their difference in the generalized join shortest queue with two waiting lines.

The decay rates for this queue have been studied in [3] and [6], but they are obtained only for certain limited cases, e.g., under a so called strongly pooled condition. We completely answer to this problem for the weak decay rates, and give weaker sufficient conditions for the exactly geometric decay rates (Theorem 1.5 and Corollary 1.2). In particular, it turns out that the strongly pooled condition still plays an important role for finding the decay rate for the minimum of two queues, which may not be the square of the total traffic intensity in general.

The two sided *DQBD* is a special case of the double sided *QBD* introduced in [3] since the latter allows the background process to be a general Markov chain. The exactly geometric decays are studied in [3], but only sufficient conditions are obtained. Furthermore, those sufficient conditions require the stationary probabilities at the boundaries, i.e., at level 0, so they are not easy to verify. Not for the two sided *DQBD* but for the *QBD*, Miyazawa [1] completely solves the decay rate problem recently, developing the ideas in [2].

We here extend this approach in [1]. Thus, many arguments are parallel to those in [1]. Namely, the approach heavily depends on the *QBD* structure and the Wiener Hopf factorization for the Markov additive process that generate the *QBD* process, and the key idea is to formulate the decay rate problem as a multidimensional optimization problem. However, the level and background states are not symmetric in the two sided *DQBD* while they are symmetric in the *QBD*. So, we need some further effort to get the decay rates, which is a main contribution of this chapter for a general *QBD* model.

For the join shortest queue and its generalized versions, the decay rate problem has been widely studied in the literature. One possible approach is to use the large deviation principle. Puhalskii and Vladimirov [7] recently obtained the weak decay rates as the solutions of the variational problem for a much more general class of the generalized join shortest queue with an arbitrary number of parallel queues. However, this variational problem is very hard to not only analytically but also numerically solve even for the case of two queues.

Another approach is either to use the random walk structure or the *QBD* formulation. For example, Foley and McDonald [6] took the former formulation while Li, Miyazawa and Zhao [3] took the latter formulation. An interesting sufficient condition, i.e., so called strongly pooled condition, is found in [6]. However, those papers mainly consider the decay rate under this limited condition for the case of the two queues. So far, the decay rate problem has not been well answered for the generalized join shortest queue. In this chapter, we completely solve this problem for the case of the two queues (Theorem 1.5 and Corollary 1.2). For simpler arrival processes, there are many other studies on the join shortest queues and the decay rate problem has been relatively well answered (see references in [3], [6]).

This chapter is made up by seven sections. In [Sect. 1.2](#), we introduce the two sided *DQBD* process formally, and consider its basic property, particularly on the rate matrices for representing the stationary distribution in a matrix geometric form. In [Sect. 1.3](#), we characterize the set of positive eigenvectors of the rate matrices using the moment generating functions of the transition kernels insides and on the boundaries. The weak decay rates are completely answered in [Sect. 1.4](#). We also give sufficient conditions for those decay rates to be exactly geometric. In [Sect. 1.5](#), we consider the generalized join the shortest queue with two queues, and answer to the decay rate problems. We finally give some remarks on the existence results in [Sect. 1.6](#). Conclusions are drawn in [Sect. 1.7](#).

1.2 Two Sided DQBD Process

Let $\{(L_{1t}, L_{2t}); t = 0, 1, \dots\}$ be a two dimensional Markov chain taking values in $S \equiv \mathbb{Z} \times \mathbb{Z}_+$, where \mathbb{Z} is the set of all integers and $\mathbb{Z}_+ = \{\ell \in \mathbb{Z}; \ell \geq 0\}$, with the following transition probabilities (see Fig. 1.1).

$$P(L_{1(t+1)} = i', L_{2(t+1)} = j' | L_{1t} = i, L_{2t} = j) = \begin{cases} p_{(i'-i)(j'-j)}^+, & i \geq 1, j \geq 1, i' - i, j' - j = 0, \pm 1 \\ p_{(i'-i)(j'-j)}^-, & i \leq -1, j \geq 1, i' - i, j' - j = 0, \pm 1 \\ p_{(i'-i)j'}^{(1+)}, & i \geq 1, j = 0, i' - i = 0, \pm 1, j' = 0, 1 \\ p_{(i'-i)j'}^{(1-)}, & i \leq -1, j = 0, i' - i = 0, \pm 1, j' = 0, 1 \\ p_{i'(j'-j)}^{(2)}, & i = 0, j \geq 1, i' = 0, 1, j' - j = 0, \pm 1 \\ p_{i'j'}^{(0)}, & i = j = 0, i' = 0, \pm 1, j' = 0, 1 \\ 0, & \text{otherwise,} \end{cases}$$

where $\sum_{i,j} p_{ij} = \sum_{i,j} p_{ij}^{(k)} = 1$ for $k = 0, \pm 1, \pm 2$. Thus, $\{(L_{1t}, L_{2t})\}$ is a skip free random walk in all directions, and reflected at the boundary $\partial S_1 \equiv \{(i, j) \in S; j = 0\}$ and has discontinuous statistics at $\partial S_2 \equiv \{(i, j) \in S; i = 0\}$.

We first take L_{1t} as level, and L_{2t} as background state, and refer to this Markov chain as a discrete-time two sided DQBD (doubly quasi-birth-and-death) process. In the random walk terminology, this process is two dimensional reflected random walk on the half space $\{(m, n) \in \mathbb{Z}^2; n \geq 0\}$ with discontinuous statistics at the boundaries where either one of components vanishes. We also note that this model is a special case of the double sided QBD in [3] whose background process is not necessary to be birth-and-death.

To present the transition probability matrix of this Markov chain, we first introduce the following matrices. For $k = 0, \pm 1$ and $s = \pm$,

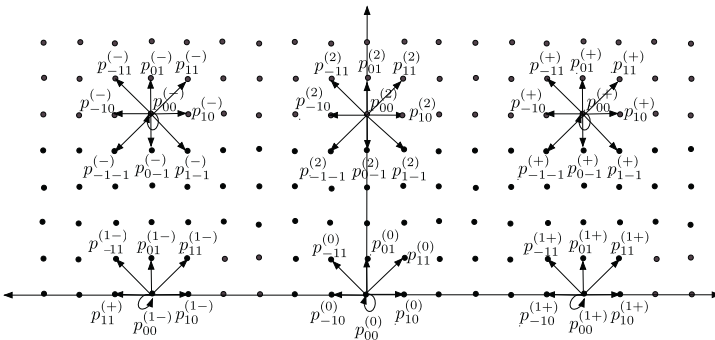


Fig. 1.1 State transitions for the two sided DQBD process.

$$A_k^{(s)} = \begin{pmatrix} P_{k0}^{(1s)} & P_{k1}^{(1s)} & 0 & \cdots \\ P_{k(-1)}^{(s)} & P_{k0}^{(s)} & P_{k1}^{(s)} & 0 & \cdots \\ 0 & P_{k(-1)}^{(s)} & P_{k0}^{(s)} & P_{k1}^{(s)} & 0 & \cdots \\ 0 & 0 & P_{k(-1)}^{(s)} & P_{k0}^{(s)} & P_{k1}^{(s)} & 0 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix},$$

and for $k = 0, \pm 1$,

$$B_k^{(1)} = \begin{pmatrix} P_{k0}^{(0)} & P_{k1}^{(0)} & 0 & \cdots \\ P_{k(-1)}^{(2)} & P_{k0}^{(2)} & P_{k1}^{(2)} & 0 & \cdots \\ 0 & P_{k(-1)}^{(2)} & P_{k0}^{(2)} & P_{k1}^{(2)} & 0 & \cdots \\ 0 & 0 & P_{k(-1)}^{(2)} & P_{k0}^{(2)} & P_{k1}^{(2)} & 0 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix}.$$

Then, the two sided DQBD has the following tridiagonal transition matrix $P^{(1)}$.

$$P^{(1)} = \begin{pmatrix} \ddots & \ddots & \ddots & \ddots & \ddots & \cdots \\ \cdots & 0 & A_{-1}^{(-)} & A_0^{(-)} & A_1^{(-)} & 0 & \cdots \\ \cdots & 0 & A_{-1}^{(-)} & A_0^{(-)} & A_1^{(-)} & 0 & \cdots \\ \cdots & 0 & B_{-1}^{(1)} & B_0^{(1)} & B_1^{(1)} & 0 & \cdots \\ \cdots & 0 & A_{-1}^{(+)} & A_0^{(+)} & A_1^{(+)} & 0 & \cdots \\ \cdots & 0 & A_{-1}^{(+)} & A_0^{(+)} & A_1^{(+)} & 0 & \cdots \\ \cdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix}.$$

Throughout this chapter, we assume that $P^{(1)}$ is irreducible and aperiodic, and positive recurrent. The unique stationary distribution of P is denoted by probability row vector:

$$\mathbf{v} = (\dots, \mathbf{v}_{-1}, \mathbf{v}_0, \mathbf{v}_1, \dots),$$

where \mathbf{v}_n for $n \in \mathbb{Z}$ are row vectors for background states in level n . We also write \mathbf{v} as $\{\mathbf{v}_{ij}; i \in \mathbb{Z}, j \in \mathbb{Z}_+\}$. We assume that

- (i) For each $s = \pm, 2$, $A^{(s)} \equiv A_{-1}^{(s)} + A_0^{(s)} + A_1^{(s)}$ is irreducible and aperiodic;
- (ii) For each $s = \pm, 2$, Markov additive process driven by kernel $\{A_n^{(s)}; n = 0, \pm 1\}$ is 1-arithmetic in the sense that for every pair $(i, j) \in S_1 \times S_1$, the greatest common divisor of $\{n \in \mathbb{Z}; A_n^{(s)}(i, j) > 0\}$ is one, where \mathbb{Z} is the set of all integers (see, e.g., [8]).

Remark 1.1. The irreducibility of $A^{(s)}$ in (i) is satisfied by many applications, but it is stronger than the irreducibility of P . Our arguments in this chapter can be modified

so as to be valid without that irreducibility, and the same results are obtained. However, proofs becomes complicated just because we need to consider each case separately depending on the irreducibility or the non irreducibility. So, we here do not consider the non irreducible case, which will be detailed in a technical note.

It is well-known that, for each $s = \pm$, there exists a nonnegative matrix $R^{(s)}$ uniquely determined as a minimal nonnegative solution of the matrix equation:

$$R^{(-)} = A_{-1}^{(-)} + R^{(-)}A_0^{(-)} + (R^{(-)})^2A_1^{(-)}, \quad (1.1)$$

$$R^{(+)} = (R^{(+)})^2A_{-1}^{(+)} + R^{(+)}A_0^{(+)} + A_1^{(+)}, \quad (1.2)$$

and the stationary distribution has the following matrix geometric form.

$$v_n = \begin{cases} v_1 (R^{(+)})^{n-1}, & n \geq 1 \\ v_{-1} (R^{(-)})^{-n-1}, & n \leq -1. \end{cases} \quad (1.3)$$

Note that $R^{(s)}$ may not be irreducible, but has a single irreducible class due to (i) and (ii).

We also consider the case that L_2 is taken as level. In this case, the transition matrix is denoted by $P^{(2)}$, and given by

$$P^{(2)} = \begin{pmatrix} B_0^{(2)} & B_1^{(2)} & 0 & \dots \\ A_{-1}^{(2)} & A_0^{(2)} & A_1^{(2)} & 0 & \dots \\ 0 & A_{-1}^{(2)} & A_0^{(2)} & A_1^{(2)} & 0 & \dots \\ 0 & 0 & A_{-1}^{(2)} & A_0^{(2)} & A_1^{(2)} & 0 & \dots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix},$$

where, for $k = 0, \pm 1$,

$$A_k^{(2)} = \begin{pmatrix} \ddots & \ddots & \ddots & \ddots & \ddots & \dots \\ \dots & 0 & p_{(-1)k}^{(-)} & p_{0k}^{(-)} & p_{1k}^{(-)} & 0 & \dots \\ \dots & 0 & p_{(-1)k}^{(-)} & p_{0k}^{(-)} & p_{1k}^{(-)} & 0 & \dots \\ \dots & 0 & p_{-1k}^{(2)} & p_{0k}^{(2)} & p_{1k}^{(2)} & 0 & \dots \\ \dots & 0 & p_{-1k}^{(+)} & p_{0k}^{(+)} & p_{1k}^{(+)} & 0 & \dots \\ \dots & 0 & p_{(-1)k}^{(+)} & p_{0k}^{(+)} & p_{1k}^{(+)} & 0 & \dots \\ \dots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix},$$

and for $k = 0, 1$,

$$B_k^{(2)} = \begin{pmatrix} \ddots & \ddots & \ddots & \ddots & \ddots & \dots \\ \dots & 0 & p_{(-1)k}^{(1-)} & p_{0k}^{(1-)} & p_{1k}^{(1-)} & 0 & \dots \\ \dots & \dots & 0 & p_{(-1)k}^{(1-)} & p_{0k}^{(1-)} & p_{1k}^{(1-)} & 0 & \dots \\ \dots & \dots & \dots & 0 & p_{-1k}^{(2)} & p_{0k}^{(2)} & p_{1k}^{(2)} & 0 & \dots \\ \dots & \dots & \dots & \dots & 0 & p_{(-1)k}^{(1+)} & p_{0k}^{(1+)} & p_{1k}^{(1+)} & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & 0 & p_{(-1)k}^{(1+)} & p_{0k}^{(1+)} & p_{1k}^{(1+)} & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix}.$$

In this case, the stationary distribution $\mathbf{v} = \{v_{ij}\}$ is partitioned as

$$\mathbf{v} = \left(\mathbf{v}_0^{(2)}, \mathbf{v}_1^{(2)}, \dots \right),$$

where $\mathbf{v}_n^{(2)} = \{v_{in}; i \in \mathbb{Z}\}$. Viewing L_{1t} as the background process, we have the standard process. Then, as is well known, there exists a minimal nonnegative solution $R^{(2)}$ of

$$R^{(2)} = (R^{(2)})^2 A_{-1}^{(2)} + R^{(2)} A_0^{(2)} + A_1^{(2)}, \quad (1.4)$$

and the stationary distribution \mathbf{v} has the following form:

$$\mathbf{v}_n^{(2)} = \mathbf{v}_1^{(2)} (R^{(2)})^{n-1}, \quad n \geq 1. \quad (1.5)$$

We are interested in the geometric decay behaviors of the stationary vector \mathbf{v}_n as $n \rightarrow \pm\infty$ and $\mathbf{v}_n^{(2)}$ as $n \rightarrow \infty$. We are interested in two different types of asymptotics. If there are constant $\alpha_+ > 1$ and constant positive vector \mathbf{c}_+ such that

$$\lim_{n \rightarrow \infty} \alpha_+^n \mathbf{v}_n = \mathbf{c}_+,$$

then \mathbf{v}_n is said to asymptotically have exactly geometric decay rate α_+^{-1} as $n \rightarrow \infty$. Another decay rate is of logarithmic type, which is defined through

$$\log r_+(i) = \lim_{n \rightarrow \infty} \frac{1}{n} \log v_{ni}, \quad i \in \mathbb{Z}_+, \quad (1.6)$$

where $r_+(i) \leq 1$. If $r_+(i)$ does not depend on i , we write it as r_+ . In this case, \mathbf{v}_n is said to asymptotically have weak geometric decay rate r_+ . Those decay rates are also defined for \mathbf{v}_n as $n \rightarrow -\infty$ and for $\mathbf{v}_n^{(2)}$ as $n \rightarrow \infty$, which are denoted by r_- and r_2 , respectively. Since those decay rates may not exist, we also use the following notation:

$$\log \underline{r}_+(i) = \liminf_{n \rightarrow \infty} \frac{1}{n} \log v_{ni}, \quad \log \bar{r}_+(i) = \limsup_{n \rightarrow \infty} \frac{1}{n} \log v_{ni}, \quad i \in \mathbb{Z}_+.$$

Similarly, $r_s(i)$ and $\bar{r}_s(i)$ are defined for $s = -, 2$. These decay rates are referred to as the weak lower and weak upper decay rates, respectively.

It is noticed that $\bar{r}_+(i)$ in (1.6) is bounded as

$$r_+(i)^{-1} \leq \sup \left\{ z \geq 1; \sum_{n=0}^{\infty} z^n v_{ni} < \infty \right\}, \quad i \in \mathbb{Z}_+.$$

Then, from (1.3) and (1.5), it might be expected that the weak decay rate r_+^{-1} is obtained as the reciprocal of the convergence parameter $c_p(R^{(+)})$ of $R^{(+)}$, which is defined as

$$c_p(R^{(+)}) = \sup \left\{ z \geq 0; \sum_{n=0}^{\infty} z^n (R^{(+)})^n < \infty \right\}.$$

This is true under certain situations, but generally not true. In general, we only have the following lower bounds for the decay rates from this information.

Lemma 1.1. The decay rates are bounded below by the corresponding convergence parameters of the rate matrices. That is, we have

$$r_s(i) \geq c_p(R^{(s)})^{-1}, \quad s = \pm, 2, \quad i \in \mathbb{Z}_s,$$

where $\mathbb{Z}_2 = \mathbb{Z}$.

The proof of this lemma is exactly the same as Lemma 2.1 of [1], so it is omitted. This lemma just gives the lower bounds, but it turns out that they are very useful to identify the decay rates as well as to prove their existence.

We next prepare some useful facts for the convergence parameters.

Proposition 1.1 (Theorem 6.3 of [9]). For a nonnegative square matrix T , let \mathbf{X} be the set of all nonnegative and nonzero row vectors whose size is the same as that of T . Then we have

$$c_p(T) = \sup \{ z \geq 0; z\mathbf{x}T \leq \mathbf{x}, \mathbf{x} \in \mathbf{X} \}.$$

We will consider all eigenvectors of $R^{(s)}$, $s = \pm, 2$, to find the decay rate. For this, we use the Markov additive process generated by $\{A_k^{(s)}; k = 0, \pm 1\}$. Note that (1.1) and (1.2) and the corresponding equations of $R^{(-)}$ and $R^{(2)}$ are equivalent to

$$I - A_*^{(s)}(z^{u(s)}) = (I - zR^{(s)})(I - G_*^{(s)}(z)), \quad z \neq 0, s = \pm, 2, \quad (1.7)$$

where $u(s) = -1$ for $s = -$, $u(s) = 1$ for $s = +, 2$, and $A_*(z)$ and $G_*^{(s)}(z)$ are defined as

$$A_*^{(s)}(z) = z^{-1}A_{-1}^{(s)} + A_0^{(s)} + zA_1^{(s)}, \quad G_*^{(s)}(z) = A_0^{(s)} + R^{(s)}A_{-1}^{(s)} + z^{-1}A_{-1}^{(s)}.$$

1.3 Eigenvectors of Rate Matrices

If we take L_1 as level, then the background process $\{L_{2t}\}$ is the birth and death process in each half lie $[1, \infty)$ or $(-\infty, -1]$. Hence, this case is easier, so we consider $R^{(s)}$ with $s = \pm$ first. In what follows, we use the following notations for $s = \pm, 2$.

$$\begin{aligned}\mathcal{V}_R^{(s)} &= \left\{ (z, \mathbf{x}); z\mathbf{x}R^{(s)} = \mathbf{x}, z \geq 1, \mathbf{x} \in \mathbf{X}^{(s)} \right\}, \\ \mathcal{V}_A^{(s)} &= \left\{ (z, \mathbf{x}); \mathbf{x}A^{(s)}(z) = \mathbf{x}, z \geq 1, \mathbf{x} \in \mathbf{X}^{(s)} \right\}.\end{aligned}$$

We first note the following facts, which are easily concluded by the Wiener Hopf factorization.

Lemma 1.3. Let s be either one of $-$, $+$ or 2 . For $z > 1$, $(z, \mathbf{x}) \in \mathcal{V}_R^{(s)}$ if and only if $(z, \mathbf{x}) \in \mathcal{V}_A^{(s)}$. If there is no (z, \mathbf{x}) in $\mathcal{V}_A^{(s)}$ with $z > 1$, then $c_p(R^{(s)}) = 1$.

Then, the following result is immediate from Theorem 3.1 in [1].

Theorem 1.1. Let $\mathcal{D}_1^{(-)}$ denote the subset of all $(-\theta_1, \theta_2)$ in \mathbb{R}^2 such that

$$\mathbb{E} \left[e^{\theta_1 X_1^{(-)} + \theta_2 X_2^{(-)}} \right] = 1, \quad (1.9)$$

$$\begin{aligned}\varphi_0^{(1-)}(\theta_1) + \varphi_1^{(1-)}(\theta_1) e^{\theta_2} &\leq 1, \\ \theta_1 &\leq 0, \theta_2 \in \mathbb{R},\end{aligned} \quad (1.10)$$

where $\varphi_i^{(1-)}(\theta_1) = \mathbb{E} \left[e^{\theta_1 X_1^{(1-)}}; X_2^{(1-)} = j \right]$ for $j = 0, 1$. Similarly, let $\mathcal{D}_1^{(+)}$ denote the subset of all (θ_1, θ_2) in \mathbb{R}^2 such that

$$\mathbb{E} \left[e^{\theta_1 X_1^{(+)} + \theta_2 X_2^{(+)}} \right] = 1, \quad (1.11)$$

$$\begin{aligned}\varphi_0^{(1+)}(\theta_1) + \varphi_1^{(1+)}(\theta_1) e^{\theta_2} &\leq 1, \\ \theta_1 &\geq 0, \theta_2 \in \mathbb{R},\end{aligned} \quad (1.12)$$

where $\varphi_i^{(1+)}(\theta_1) = \mathbb{E} \left[e^{\theta_1 X_1^{(1+)}}; X_2^{(1+)} = j \right]$ for $j = 0, 1$. Then, for each $s = \pm$, there exists a $(z, \mathbf{x}) \in \mathcal{V}_A^{(s)}$ if and only if there exists a $(\theta_1, \theta_2) \in \mathcal{D}_1^{(s)}$. Furthermore, we have the following facts.

(1a) For this (θ_1, θ_2) , $(z, \mathbf{x}) \in \mathcal{V}_A^{(-)}$ (res., $\mathcal{V}_A^{(+)}$) is given by $z = e^{\theta_1}$ and $\mathbf{x} = \{x_n\}$:

$$x_n = \begin{cases} c_1 e^{-\underline{\theta}_2(n-1)} + c_2 e^{-\bar{\theta}_2(n-1)}, & \underline{\theta}_2 \neq \bar{\theta}_2, \\ (c'_1 + c'_2(n-1)) e^{-\underline{\theta}_2(n-1)}, & \underline{\theta}_2 = \bar{\theta}_2, \end{cases} \quad n \geq 1, \quad (1.13)$$

where $\underline{\theta}_2, \bar{\theta}_2$ are the two solutions of (1.9) (res., (1.11)) for the given θ_1 such that $\underline{\theta}_2 \leq \bar{\theta}_2$, and c_i, c'_i are nonnegative constants satisfying $c_1 + c_2 \neq 0$ and $c'_1 +$

$c'_2 \neq 0$. Furthermore, both of c_1 and c_2 are positive only if the strict inequality holds in (1.9) (res., (1.11)).

(1b) The convergence parameter $c_p(R^{(s)})$ is obtained as the supremum of e^{θ_1} over $\mathcal{D}_1^{(s)}$ for each $s = \pm$.

Similarly to Theorem 1.1, we can prove the following theorem for $R^{(2)}$. Since this result is the core of our arguments, we give its detailed proof in Appendix A.

Theorem 1.2. Let \mathcal{D}_2 denote the subset of all $(-\eta_1^{(-)}, \eta_1^{(+)}, \eta_2)$ in \mathbb{R}^3 such that

$$\mathbb{E} \left[e^{\eta_1^{(-)} X_1^{(-)} + \eta_2 X_2^{(-)}} \right] = 1, \quad (1.14)$$

$$\mathbb{E} \left[e^{\eta_1^{(+)} X_1^{(+)} + \eta_2 X_2^{(+)}} \right] = 1, \quad (1.15)$$

$$\varphi_{-1}^{(2)}(\eta_2) e^{-\eta_1^{(-)}} + \varphi_0^{(2)}(\eta_2) + \varphi_1^{(2)}(\eta_2) e^{\eta_1^{(+)}} \leq 1, \quad (1.16)$$

$$\eta_2 \geq 0, \eta_1^{(-)}, \eta_1^{(+)} \in \mathbb{R},$$

where $\varphi_i^{(2)}(\eta_2) = \mathbb{E} \left[e^{\eta_2 X_2^{(2)}}; X_1^{(2)} = i \right]$ for $i = 0, \pm 1$. Then, there exists a $(z, \mathbf{x}) \in \mathcal{V}_A^{(2)}$ if and only if there exists a $(-\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2$. Furthermore, we have the following facts.

(2a) For this $(-\eta_1^{(-)}, \eta_1^{(+)}, \eta_2), (z, \mathbf{x}) \in \mathcal{V}_A^{(2)}$ is given by $z = e^{\eta_2}$ and $\mathbf{x} = \{x_n\}$:

$$x_n^{(s)} = \begin{cases} c_1^{(s)} e^{-\underline{\eta}_1^{(s)}(n-1)} + c_2^{(s)} e^{-\bar{\eta}_1^{(s)}(n-1)}, & \underline{\eta}_1^{(s)} \neq \bar{\eta}_1^{(s)}, \\ \left(d_1^{(s)} + d_2^{(s)} |n-1| \right) e^{-\underline{\eta}_1^{(s)}(n-1)}, & \underline{\eta}_1^{(s)} = \bar{\eta}_1^{(s)} \end{cases} \quad n \geq 1, s = \pm, \quad (1.17)$$

where $\underline{\eta}_1^{(-)}, \bar{\eta}_1^{(-)}$ (res., $\underline{\eta}_1^{(+)}, \bar{\eta}_1^{(+)}$) are the two solutions of (1.14) (res., (1.15)) for the given η_2 such that $\underline{\eta}_1^{(-)} \leq \bar{\eta}_1^{(-)}$ (res., $\underline{\eta}_1^{(+)} \leq \bar{\eta}_1^{(+)}$), and for each $s = \pm$, $c_i^{(s)}, d_i^{(s)}$ are nonnegative constants satisfying $c_1^{(s)} + c_2^{(s)} \neq 0$ and $d_1^{(s)} + d_2^{(s)} \neq 0$. Furthermore, both of $c_1^{(s)}$ and $c_2^{(s)}$ are positive only if the strict inequality holds in (1.16).

(2b) The convergence parameter $c_p(R^{(2)})$ is obtained as the supremum of e^{η_2} over \mathcal{D}_2 .

For convenience, we also introduce the following projections of \mathcal{D}_2 , which will be used in Lemma 1.7.

$$\begin{aligned} \mathcal{D}_2^{(-)} &= \{(\eta_1^{(-)}, \eta_2); (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2\}, \\ \mathcal{D}_2^{(+)} &= \{(\eta_1^{(+)}, \eta_2); (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2\}. \end{aligned}$$

An important observation in these theorems is that z satisfying $(z, \mathbf{x}) \in \mathcal{V}(R^{(s)})$ can be found through θ_1 or η_2 in sets $\mathcal{D}_1^{(-)}, \mathcal{D}_1^{(+)}$ and \mathcal{D}_2 , which are in the boundary

of convex sets. Furthermore, $\mathcal{D}_1^{(s)}$, $\mathcal{D}_2^{(2)}$ and \mathcal{D}_2 are compact and connected sets for $s = \pm$. This observation is expected to extend Corollary 3.1 of [1] for the two sided QBD process. However, we have to check the two sided version of Proposition 3.1 of [1]. That is, we need the following lemmas. For convenience, we denote the set of non-positive integers by \mathbb{Z}_- .

Lemma 1.4. For $s = \pm$, if there exist a positive vector $\mathbf{x}^{(s)} = \{x_n^{(s)}; n \in \mathbb{Z}_s\}$ such that $(\alpha_s, \mathbf{x}^{(s)}) \in \mathcal{V}_A^{(s)}$ and some finite $\underline{d}_s(\mathbf{x}), \bar{d}_s(\mathbf{x}) \geq 0$ such that

$$\liminf_{n \rightarrow \infty} \frac{V_{sn}}{x_n} = \underline{d}_s(\mathbf{x}^{(s)}), \quad \limsup_{n \rightarrow \infty} \frac{V_{sn}}{x_n} = \bar{d}_s(\mathbf{x}^{(s)}),$$

then, for any nonnegative column vector $\mathbf{u}^{(s)}$ satisfying $\mathbf{x}^{(s)} \mathbf{u}^{(s)} < \infty$, there are nonnegative and finite $\underline{d}_s^\dagger(\mathbf{x}^{(s)})$ and $\bar{d}_s^\dagger(\mathbf{x}^{(s)})$ such that

$$\begin{aligned} \alpha_s \underline{d}_s^\dagger(\mathbf{x}^{(s)}) \mathbf{x}^{(s)} \mathbf{u}^{(s)} &\leq \liminf_{n \rightarrow s\infty} \alpha_s^{|n|} v_n \mathbf{u}^{(s)} \\ &\leq \limsup_{n \rightarrow s\infty} \alpha_s^{|n|} v_n \mathbf{u}^{(s)} \leq \alpha_s \bar{d}_s^\dagger(\mathbf{x}^{(s)}) \mathbf{x}^{(s)} \mathbf{u}^{(s)}. \end{aligned} \quad (1.18)$$

In particular, if $\underline{d}_s^\dagger(\mathbf{x}^{(s)}) = \bar{d}_s^\dagger(\mathbf{x}^{(s)})$ and $0 \leq d_s^\dagger \equiv \underline{d}_s^\dagger(\mathbf{x}^{(s)}) < \infty$, then

$$\lim_{n \rightarrow s\infty} \alpha_s^n v_n \mathbf{u}^{(s)} = \alpha_s d_s^\dagger \mathbf{x}^{(s)} \mathbf{u}^{(s)}. \quad (1.19)$$

That is, $v_n \mathbf{u}^{(s)}$ decays geometrically with rate α_s^{-1} as $n \rightarrow s\infty$.

Lemma 1.5. If there exist a positive vector $\mathbf{x} = \{x_n; n \in \mathbb{Z}\}$ such that $(\alpha, \mathbf{x}) \in \mathcal{V}_A^{(2)}$ and some finite $\underline{d}^-(\mathbf{x}), \bar{d}^-(\mathbf{x}), \underline{d}^+(\mathbf{x}), \bar{d}^+(\mathbf{x}) \geq 0$ such that

$$\begin{aligned} \liminf_{n \rightarrow -\infty} \frac{V_{n1}}{x_n} &= \underline{d}^-(\mathbf{x}), & \limsup_{n \rightarrow -\infty} \frac{V_{n1}}{x_n} &= \bar{d}^-(\mathbf{x}), \\ \liminf_{n \rightarrow +\infty} \frac{V_{n1}}{x_n} &= \underline{d}^+(\mathbf{x}), & \limsup_{n \rightarrow +\infty} \frac{V_{n1}}{x_n} &= \bar{d}^+(\mathbf{x}), \end{aligned}$$

then, for any nonnegative column vector \mathbf{u} satisfying $\mathbf{x} \mathbf{u} < \infty$, there are nonnegative and finite $\underline{d}^\dagger(\mathbf{x})$ and $\bar{d}^\dagger(\mathbf{x})$ such that

$$\alpha \underline{d}^\dagger(\mathbf{x}) \mathbf{x} \mathbf{u} \leq \liminf_{n \rightarrow \infty} \alpha^n v_n^{(2)} \mathbf{u} \leq \limsup_{n \rightarrow \infty} \alpha^n v_n^{(2)} \mathbf{u} \leq \alpha \bar{d}^\dagger(\mathbf{x}) \mathbf{x} \mathbf{u}. \quad (1.20)$$

In particular, if $\underline{d}^\dagger(\mathbf{x}) = \bar{d}^\dagger(\mathbf{x})$ and $0 \leq d^\dagger \equiv \underline{d}^\dagger(\mathbf{x}) < \infty$, then

$$\lim_{n \rightarrow \infty} \alpha^n v_n^{(2)} \mathbf{u} = \alpha d^\dagger \mathbf{x} \mathbf{u}. \quad (1.21)$$

That is, $v_n^{(2)} \mathbf{u}$ decays geometrically with rate α^{-1} as n goes to infinity.

Since this lemma can be proved in a similar way to Proposition 3.1 of [1], we omit its proof. For each $n \geq 0$, let

$$\mathbf{v}_{-n}^{(-)} = \{\mathbf{v}_{(-n)k}; k \geq 0\}, \quad \mathbf{v}_n^{(+)} = \{\mathbf{v}_{nk}; k \geq 0\}, \quad \mathbf{v}_n^{(2)} = \{\mathbf{v}_{kn}; k \in \mathbb{Z}\}.$$

Then, the next corollary follows from Theorem 1.1, Theorem 1.2 and Lemmas 1.4 and 1.5 similarly to Corollary 3.1 of [1].

Corollary 1.1. Define β_s for $s \pm, 2$ as

$$\begin{aligned} \beta_- &= \sup \left\{ \theta_1; \limsup_{n \rightarrow \infty} \mathbf{v}_{(-1)n} e^{\theta_2 n} < \infty, (\theta_1, \theta_2) \in \mathcal{D}_1^{(-)} \right\}, \\ \beta_+ &= \sup \left\{ \theta_1; \limsup_{n \rightarrow \infty} \mathbf{v}_{1n} e^{\theta_2 n} < \infty, (\theta_1, \theta_2) \in \mathcal{D}_1^{(+)} \right\}, \\ \beta_2 &= \sup \left\{ \eta_2; \limsup_{n \rightarrow \infty} \mathbf{v}_{(-n)1} e^{\eta_1^{(-)} n} < \infty, \right. \\ &\quad \left. \limsup_{n \rightarrow \infty} \mathbf{v}_{n1} e^{\eta_1^{(+)} n} < \infty, (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2 \right\}. \end{aligned}$$

Then, the weak upper decay rates $\bar{r}_-(i)$, $\bar{r}_+(i)$ and $\bar{r}_2(j)$ of \mathbf{v}_{-ni} , \mathbf{v}_{ni} and \mathbf{v}_{jn} , respectively, as $n \rightarrow \infty$ are uniformly bounded by $e^{-\beta_-}$, $e^{-\beta_+}$ and $e^{-\beta_2}$. In particular, for each $s = \pm, 2$, if $\beta_s = \log c_p(R^{(s)})$, then the weak decay rate r_s exists and $r_s = e^{-\beta_s}$. Furthermore, if the asymptotic decay of \mathbf{v}_{1n} , $\mathbf{v}_{(-1)n}$ or \mathbf{v}_{n1} and $\mathbf{v}_{-n(-1)}$ is exactly geometric as $n \rightarrow \infty$, then the corresponding stationary level distribution asymptotically decays in the exactly geometric form.

1.4 Answers to Decay Rate Problem

We are now in a position to answer to the decay rate problem. Since $\mathcal{D}_1^{(s)}$ for $s = \pm$ and \mathcal{D}_2 are compact sets, we can define, for $s = \pm$,

$$\begin{aligned} \theta_1^{(sc)} &= \max\{\theta_1; (\theta_1, \theta_2) \in \mathcal{D}_1^{(s)}\}, & \theta_2^{(sc)} &= \min\{\theta_2; (\theta_1^{(sc)}, \theta_2) \in \mathcal{D}_1^{(s)}\}, \\ \eta_2^{(c)} &= \max\{\eta_2; (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2\}, \\ \eta_1^{(sc)} &= \max\{\eta_1^{(s)}; (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2^{(c)}) \in \mathcal{D}_2\}. \end{aligned}$$

Note that $\theta_1^{(sc)} = \log c_p(R^{(s)})$ for $s = \pm$ and $\eta_2^{(c)} = \log c_p(R^{(2)})$. Furthermore, $(\eta_1^{(-)}, \eta_1^{(+)}, \eta_2^{(c)})$ and $(\theta_1^{(sc)}, \theta_2^{(sc)})$ are in \mathcal{D}_2 and $\mathcal{D}_1^{(s)}$ for $s = \pm$, respectively.

Similarly to Theorem 4.1 of [1], we consider the following nonlinear optimization problems. Let, for $s = \pm$,

$$\alpha_s = \sup\{e^{\theta_1^{(s)}}; \theta_2^{(s)} \leq \eta_2, \eta_1^{(-)} \leq \theta_1^{(-)}, \eta_1^{(+)} \leq \theta_1^{(+)},$$

$$(\theta_1^{(-)}, \theta_2^{(-)}) \in \mathcal{D}_1^{(-)}, (\theta_1^{(+)}, \theta_2^{(+)}) \in \mathcal{D}_1^{(+)}, (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2\}, \quad (1.22)$$

$$\alpha_2 = \sup\{e^{\eta_2}; \theta_2^{(-)} \leq \eta_2, \theta_2^{(+)} \leq \eta_2, \eta_1^{(-)} \leq \theta_1^{(-)}, \eta_1^{(+)} \leq \theta_1^{(+)},$$

$$(\theta_1^{(-)}, \theta_2^{(-)}) \in \mathcal{D}_1^{(-)}, (\theta_1^{(+)}, \theta_2^{(+)}) \in \mathcal{D}_1^{(+)}, (\eta_1^{(-)}, \eta_1^{(+)}, \eta_2) \in \mathcal{D}_2\}. \quad (1.23)$$

We can find solutions α_s for $s = \pm, 2$ in the following way.

Lemma 1.6. For the two sided DQBD process satisfying the assumptions (i) and (ii), suppose that its stationary distribution exists, which denoted by $\nu = \{v_{ij}\}$. Then, we have

$$\bar{r}_s \equiv \sup_i \bar{r}_s(i) \leq \alpha_s^{-1}, \quad s = \pm, 2. \quad (1.24)$$

Proof. We define the following functions of $u, u_-, u_+ \geq 0$.

$$f_-(u) = \sup\left\{\theta_1; \theta_2 \leq u, (\theta_1, \theta_2) \in \mathcal{D}_1^{(-)}\right\},$$

$$f_+(u) = \sup\left\{\theta_2; \theta_1 \leq u, (\theta_1, \theta_2) \in \mathcal{D}_1^{(+)}\right\},$$

$$f_2(u_-, u_+) = \sup\left\{\eta_2; \eta_1^{(-)} \leq u_-, \eta_1^{(+)} \leq u_+, (\eta_1^{(-)}, \eta_2^{(+)}, \eta_2) \in \mathcal{D}_2\right\}.$$

For convenience, let $\sigma_s = -\log \bar{r}_s$ for $s = \pm, 2$. Suppose that $0 \leq u_s \leq \sigma_s$, which implies that $\bar{r}_s(1) \leq e^{-u_s}$ and $\bar{r}_2(s) \leq e^{-u_2}$ for $s = \pm$. Then, Corollary 1.1 leads that

$$f_-(u_2) \leq \sigma_-, \quad f_+(u_2) \leq \sigma_+, \quad f_2(u_-, u_+) \leq \sigma_2. \quad (1.25)$$

We next inductively define $u_s^{(n)}$ for $n = 0, 1, \dots$ and $s = \pm, 2$ in the following way. Let $u_s^{(0)} = 0$, and

$$u_-^{(n+1)} = f_-(u_2^{(n)}), \quad u_+^{(n+1)} = f_+(u_2^{(n)}), \quad u_2^{(n+1)} = f_2(u_-^{(n+1)}, u_+^{(n+1)}).$$

Then, it is easy to see that $u_s^{(n)}$ is non decreasing in n , and satisfies (1.25) for $u_s = u_s^{(n)}$ for $s = \pm, 2$. Hence, Corollary 1.1 concludes

$$u_s^{(n)} \leq \sigma_s, \quad n = 0, 1, \dots, \quad s = \pm, 2.$$

On the other hand, from the definitions of α_s , it is easy to prove by induction that

$$u_s^{(\infty)} \equiv \lim_{n \rightarrow \infty} u_s^{(n)} \leq \log \alpha_s, \quad s = \pm, 2.$$

Then, it can be shown that the limits $u_s^{(\infty)}$ are attained in finitely many steps. The detailed proof of this can be found in the proof of Theorem 4.1 of [1]. Hence, we have

$$\log \alpha_s = \lim_{n \rightarrow \infty} u_s^{(n)} \leq \sigma_s, \quad s = \pm, 2.$$

Thus, we get (1.24). \square

For each $s = \pm$, define the following four sets of conditions.

$$\begin{aligned} (sC1) \quad & \eta_1^{(sc)} < \theta_1^{(sc)} \text{ and } \theta_2^{(sc)} < \eta_2^{(c)}, & (sC2) \quad & \eta_1^{(sc)} < \theta_1^{(sc)} \text{ and } \eta_2^{(c)} \leq \theta_2^{(sc)}, \\ (sC3) \quad & \theta_1^{(sc)} \leq \eta_1^{(sc)} \text{ and } \theta_2^{(sc)} < \eta_2^{(c)}, & (sC4) \quad & \theta_1^{(sc)} \leq \eta_1^{(sc)} \text{ and } \eta_2^{(c)} \leq \theta_2^{(sc)}. \end{aligned}$$

These conditions are exclusive and cover all the cases for each $s = \pm$. Furthermore, (sC4) is impossible since $\theta_1^{(sc)} \leq \eta_1^{(sc)}$ implies that $\eta_2^{(c)} > \theta_2^{(sc)}$ due to the convexity of the set with boundary (1.9) and (1.11). For the other three cases for each $s = \pm$, we have to consider their combinations, so nine cases in total. For convenience, we denote the condition that $(-Ci)$ and $(+Cj)$ hold by $C(i, j)$ for $i, j = 1, 2, 3$.

The next lemma shows how we can compute α_s for $s = \pm, 2$.

Lemma 1.7. Under the assumptions of Lemma 1.6, the α_- , α_+ and α_2 of (1.22) and (1.23) are obtained in either one of the following nine ways.

- (c1) If $C(1,1)$ holds, then $\alpha_- = \exp(\theta_1^{(-c)})$, $\alpha_+ = \exp(\theta_1^{(+c)})$ and $\alpha_2 = \exp(\eta_2^{(c)})$.
- (c2) If $C(1,2)$ holds, then $\alpha_- = \exp(\theta_1^{(-c)})$, $\alpha_2 = \exp(\eta_2^{(c)})$ and α_+ is the maximum value satisfying $(\log \alpha_+, \eta_2^{(c)}) \in \mathcal{D}_1^{(+)}$.
- (c3) If $C(2,1)$ holds, then $\alpha_+ = \exp(\theta_1^{(+c)})$, $\alpha_2 = \exp(\eta_2^{(c)})$ and α_- is the maximum value satisfying $(\log \alpha_-, \eta_2^{(c)}) \in \mathcal{D}_1^{(-)}$.
- (c4) If $C(1,3)$ holds, then $\alpha_+ = \exp(\theta_1^{(+c)})$, α_2 is the maximum value satisfying $(\theta_1, \log \alpha_2) \in \mathcal{D}_2^{(+)}$ with $\theta_1 \leq \theta_1^{(+c)}$, and α_- is the maximal value satisfying $(\log \alpha_-, \theta_2) \in \mathcal{D}_1^{(-)}$ with $\theta_2 \leq \alpha_2$.
- (c5) If $C(3,1)$ holds, then $\alpha_- = \exp(\theta_1^{(-c)})$, α_2 is the maximum value satisfying $(\theta_1, \log \alpha_2) \in \mathcal{D}_2^{(-)}$ with $\theta_1 \leq \theta_1^{(-c)}$, and α_+ is the maximal value satisfying $(\log \alpha_+, \theta_2) \in \mathcal{D}_1^{(+)}$ with $\theta_2 \leq \alpha_2$.
- (c6) If $C(2,2)$ holds, then $\alpha_2 = \exp(\eta_2^{(c)})$ and α_s is the maximum value satisfying $(\log \alpha_s, \eta_2^{(c)}) \in \mathcal{D}_1^{(s)}$ for $s = \pm$.
- (c7) If $C(2,3)$ holds, then $\alpha_+ = \exp(\theta_1^{(+c)})$, α_2 is the maximum value satisfying $(\theta_1, \log \alpha_2) \in \mathcal{D}_2^{(+)}$ with $\theta_1 \leq \theta_1^{(+c)}$, and α_- is the maximum value satisfying $(\log \alpha_-, \log \alpha_2) \in \mathcal{D}_1^{(-)}$.
- (c8) If $C(3,2)$ holds, then $\alpha_- = \exp(\theta_1^{(-c)})$, α_2 is the maximum value satisfying $(\theta_1, \log \alpha_2) \in \mathcal{D}_2^{(-)}$ with $\theta_1 \leq \theta_1^{(-c)}$, and α_+ is the maximum value satisfying $(\log \alpha_+, \log \alpha_2) \in \mathcal{D}_1^{(+)}$.
- (c9) If $C(3,3)$ holds, then $\alpha_- = \exp(\theta_1^{(-c)})$, $\alpha_+ = \exp(\theta_1^{(+c)})$ and α_2 is the maximum value satisfying $(\theta_1^{(-)}, \theta_1^{(+)}, \log \alpha_2) \in \mathcal{D}_2$ with $\theta_1^{(-)} \leq \theta_1^{(-c)}$ and $\theta_1^{(+)} \leq \theta_1^{(+c)}$.

This theorem can be proved in the same way as Lemma 4.2 of [1]. So, instead of proving it, we give figures to explain how those decay rates are obtained. They can be found in Figs. 1.2, 1.3 and 1.4. Since cases (c3), (c5) and (c7) are symmetric with (c2), (c4) and (c6), respectively, we omit their figures. We shall see more figures for specific examples in Sect. 1.5.

Theorem 1.3. Under the assumptions of Lemma 1.6, we have $r_s = \alpha_s^{-1}$ for $s = \pm, 2$. Namely, α_-^{-1} , α_+^{-1} and α_2^{-1} are the weak decay rates of $v_{-n}^{(-)}$, $v_n^{(+)}$ and $v_n^{(2)}$, respectively, as $n \rightarrow \infty$. Furthermore, the marginal probabilities, $v_{-n}^{(-)} \mathbf{1}$, $v_n^{(+)} \mathbf{1}$ and $v_n^{(2)} \mathbf{1}$, have the same decay rates α_-^{-1} , α_+^{-1} and α_2^{-1} , respectively, if they are less than 1, respectively.

Proof. We first consider $\bar{r}_s(1)$ for $s = \pm, 2$, which are the weak upper decay rates of v_{-n1} , v_{n1} and v_{1n} as $n \rightarrow \infty$, respectively, are obtained by (1.24). Hence, Lemmas 1.1, 1.4 and 1.5 yield

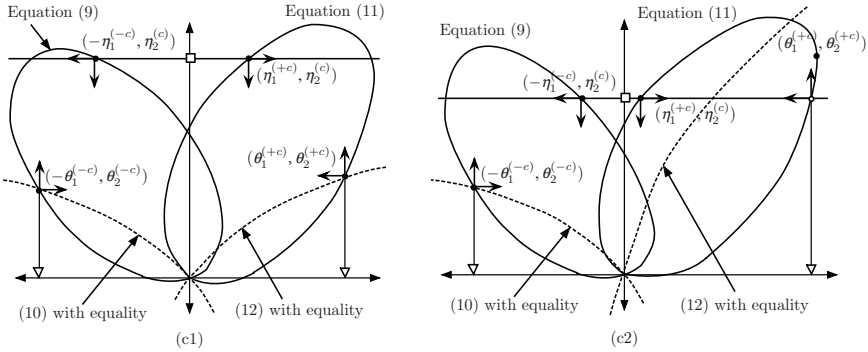


Fig. 1.2 Typical examples for (c1) and (c2).

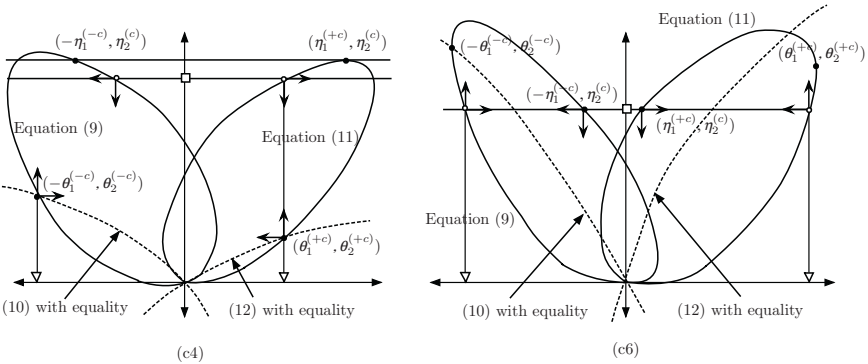


Fig. 1.3 Typical examples for (c4) and (c6).

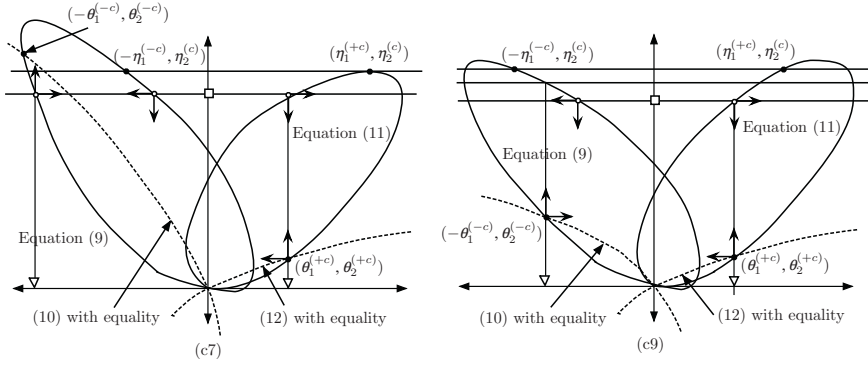


Fig. 1.4 Typical examples for (c7) and (c9).

$$c_p \left(R^{(s)} \right)^{-1} \leq r_{\pm s}(i) \leq \bar{r}_s(i) \leq \alpha_s^{-1}, \quad i \in \mathbb{Z}_+ \text{ for } s = \pm \text{ and } i \in \mathbb{Z} \text{ for } s = 2.$$

From Lemma 1.7, at least one of α_- , α_+ and α_2 agree with the corresponding convergence parameter $c_p(R^{(s)})$. Hence, we have $r_s = \alpha_s^{-1}$ at least for one s . This together with Corollary 1.1 and Lemmas 1.4 and 1.5 conclude that the same s equality must hold for the other s 's. This completes the proof. \square

We can refine the decay rates in this theorem from weak to exact ones in a similar way as Theorem 4.2 of [1] using Proposition 3.1 of [1] and Lemmas 1.4 and 1.5 for the case that the decay rates are exactly geometric. However, for the other cases, we can not directly use Theorem 5 of [10] which was used in [1] since the level or background state is two sided. Thus, we here only present the case that the exactly geometric decay occurs. We omit its proof since it is similar to Theorem 4.2 of [1].

Theorem 1.4. Under the assumptions of Theorem 1.3 with $\alpha_- > 1$, $\alpha_+ > 1$ or $\alpha_2 > 1$, let, for $s = \pm$,

$$\mathcal{D}_0^{(s)} = \left\{ (s\theta_1, \theta_2) \in \mathbb{R}^2; \mathbb{E} \left[e^{\theta_1 X_1^{(s)} + \theta_2 X_2^{(s)}} \right] = 1 \right\},$$

$$\theta_i^{s \max} = \arg \max_{(\theta_1, \theta_2) \in \mathcal{D}_0^{(s)}} \{ \theta_i \}, \quad i = 1, 2.$$

Then, we have the exactly geometric decay rates for the following cases.

- (d1) If either (-C2) or (-C3) holds, then both asymptotic decays of $\{v_{(-n)k}\}$ and $\{v_{ln}\}$ as $n \rightarrow \infty$ are exactly geometric with the decay rates α_-^{-1} and α_2^{-1} , respectively.
- (d2) If either (+C2) or (+C3) holds, then both asymptotic decays of $\{v_{nk}\}$ and $\{v_{ln}\}$ as $n \rightarrow \infty$ are exactly geometric with the decay rates α_+^{-1} and α_2^{-1} , respectively.
- (d3) If (C1) holds and if $\theta_1^{s \max} \notin \mathcal{D}_1^{(s)}$ and $\eta_1^{(c)} < \theta_1^{(sc)}$, then the asymptotic decay of $\{v_{(sn)k}\}$ ($\{v_{ln}\}$) as $n \rightarrow \infty$ is exactly geometric with the rate α_s^{-1} (α_2^{-1}) for $s = \pm$.

1.5 Generalized Join Shortest Queue

Let us apply Theorems 1.3 and 1.4 to the generalized join shortest queue which is studied in [3], [6] and explained in Sect. 1.1. We first introduce notations for this model. It has two parallel queues, numbered as queues 1 and 2. For each $i = 1, 2$, queue i serves customers in the First-Come First-Served manner with *i.i.d.* service times subject to the exponential distribution with rate μ_i . There are three exogenous Poisson arrival streams. The first and second streams go to queues 1 and 2 with the mean arrival rate λ_1 and λ_2 , respectively, while arriving customers in the third stream with the mean rate δ choose the shorter queue with tie breaking. The decay rates does not depend on the probability that customer with tie breaking choose queue 1, so we simply assume it to be $1/2$.

We are interested to see how the stationary tail probabilities of the shorter queue lengths and the difference of the two queues decay. Due to the dedicated stream to each queue, this problem is much harder than the one for the standard joining the shortest queue. Since we only consider the stationary distribution, we can formulate this continuous time model as a discrete time Markov chain. For this, we assume without loss of generality that

$$\lambda_1 + \lambda_2 + \mu_1 + \mu_2 + \delta = 1.$$

Let Q_{1t} and Q_{2t} be the queue lengths including customers being served at time $t = 0, 1, \dots$, and let $L_{1t} = Q_{2t} - Q_{1t}$ and $L_{2t} = \min(Q_{1t}, Q_{2t})$. It is not hard to see that (L_{1t}, L_{2t}) is a skip free random walk on each region $(\mathbb{Z}_+ \cup \{0\}) \times (\mathbb{Z}_+ \setminus \{0\})$ reflected at the boundary $\mathbb{Z} \times \{0\}$ and has different transitions at $\{0\} \times \mathbb{Z}_+$ (see Fig. 1.5).

Then, the transition probabilities are give by

$$\begin{aligned} p_{(-1)0}^{(-)} &= \lambda_1, & p_{(-1)(-1)}^{(-)} &= \mu_2, & p_{10}^{(-)} &= \mu_1, & p_{11}^{(-)} &= \lambda_2 + \delta, \\ p_{10}^{(+)} &= \lambda_2, & p_{1(-1)}^{(+)} &= \mu_1, & p_{(-1)0}^{(+)} &= \mu_2, & p_{(-1)1}^{(+)} &= \lambda_1 + \delta, \\ p_{10}^{(2)} &= \lambda_2 + \frac{\delta}{2}, & p_{1(-1)}^{(2)} &= \mu_1, & p_{(-1)(-1)}^{(2)} &= \mu_2, & p_{(-1)0}^{(2)} &= \lambda_1 + \frac{\delta}{2}, \end{aligned}$$

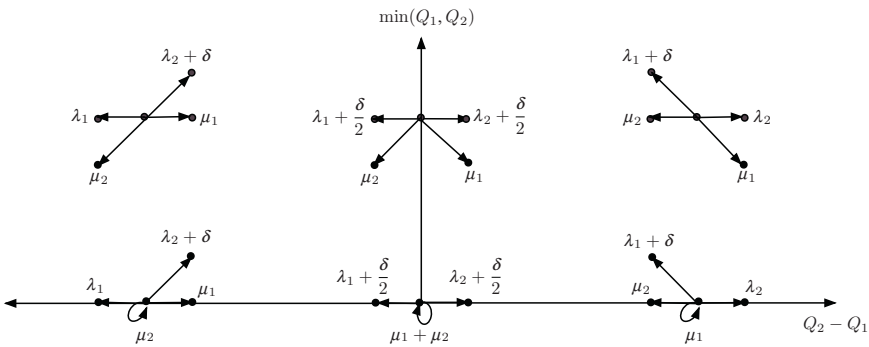


Fig. 1.5 State transitions for the generalized shortest queue.

$$\begin{aligned}
p_{(-1)0}^{(1-)} &= \lambda_1, & p_{00}^{(1-)} &= \mu_2, & p_{10}^{(1+)} &= \mu_1, & p_{11}^{(1-)} &= \lambda_2 + \delta, \\
p_{10}^{(1+)} &= \lambda_2, & p_{00}^{(1+)} &= \mu_1, & p_{(-1)0}^{(1+)} &= \mu_2, & p_{(-1)1}^{(1+)} &= \lambda_1 + \delta, \\
p_{10}^{(0)} &= \lambda_2 + \frac{\delta}{2}, & p_{00}^{(0)} &= \mu_1 + \mu_2, & p_{(-1)0}^{(0)} &= \lambda_1 + \frac{\delta}{2},
\end{aligned}$$

where all other transitions are null. To exclude obvious cases, we assume that δ, μ_1, μ_2 are all positive.

Denote traffic intensities by

$$\rho_1 = \frac{\lambda_1}{\mu_1}, \quad \rho_2 = \frac{\lambda_2}{\mu_2}, \quad \rho = \frac{\lambda_1 + \lambda_2 + \delta}{\mu_1 + \mu_2}.$$

Then, it is known that this generalized join shortest queue is stable if and only if $\rho_1 < 1, \rho_2 < 1$ and $\rho < 1$ (e.g., see [6]). This stability condition is assumed throughout this section. We will also use the following notation, which were introduced and shown to be very useful in computations in [3].

$$\gamma_1 = \mu_1 \rho^2 + \lambda_2, \quad \gamma_2 = \mu_2 \rho^2 + \lambda_1.$$

We apply Theorem 1.3 to this model. For this, we need to compute $\theta^{(-c)}, \theta^{(+c)}$ and $\eta_2^{(c)}$. In the view of Theorems 1.1 and 1.2, they are obtained if we can solve the following three sets of equations.

$$\mathbb{E} \left[e^{\theta_1 X_1^{(-)} + \theta_2 X_2^{(-)}} \right] = 1, \quad \varphi_0^{(1-)}(\theta_1) + \varphi_1^{(1-)}(\theta_1) e^{\theta_2} = 1, \quad (1.26)$$

$$\mathbb{E} \left[e^{\theta_1 X_1^{(+)} + \theta_2 X_2^{(+)}} \right] = 1, \quad \varphi_0^{(1+)}(\theta_1) + \varphi_1^{(1+)}(\theta_1) e^{\theta_2} = 1, \quad (1.27)$$

$$\begin{aligned}
\mathbb{E} \left[e^{\eta_1^{(-)} X_1^{(-)} + \eta_2 X_2^{(-)}} \right] &= 1, & \mathbb{E} \left[e^{\eta_1^{(+)} X_1^{(+)} + \eta_2 X_2^{(+)}} \right] &= 1, \\
\varphi_{-1}^{(2)}(\eta_2) e^{-\eta_1^{(-)}} + \varphi_0^{(2)}(\eta_2) + \varphi_1^{(2)}(\eta_2) e^{\eta_1^{(+)}} &= 1. & (1.28)
\end{aligned}$$

For convenience, let $z = e^{-\theta_1}$ and $\xi = e^{\theta_2}$ in (1.26). Then, we have

$$\lambda_1 z + \mu_2 z \xi^{-1} + \mu_1 z^{-1} + (\lambda_2 + \delta) z^{-1} \xi = 1, \quad (1.29)$$

$$\lambda_1 z + \mu_2 + \mu_1 z^{-1} + (\lambda_2 + \delta) z^{-1} \xi = 1. \quad (1.30)$$

Solving these equations for $z \neq 1$, we have $z = \xi = \rho_1^{-1}$. For $z = \rho_1^{-1}$, (1.29) yields $\xi = \rho_1^{-1}, \frac{\mu_2}{\lambda_2 + \delta} \rho_1^{-1}$. Note that $\rho_1^{-1} < \frac{\mu_2}{\lambda_2 + \delta} \rho_1^{-1}$ if and only if $\mu_2 > \lambda_2 + \delta$. Hence, reminding the definitions of $\theta_i^{-\max}$:

$$\theta_1^{-\max} = \max\{\log z; (1.29) \text{ holds.}\}, \quad \theta_2^{-\max} = \max\{\log \xi; (1.29) \text{ holds.}\},$$

we have

$$\left(\theta_1^{(-c)}, \theta_2^{(-c)}\right) = \begin{cases} (\log \rho_1^{-1}, \log \rho_1^{-1}), & \mu_2 > \lambda_2 + \delta \\ (\theta_1^{-\max}, \theta_2^{-\max}), & \mu_2 \leq \lambda_2 + \delta. \end{cases} \quad (1.31)$$

It is also notable that $\theta_1^{-\max} \geq \log \rho_1^{-1}$, so we always have that $\theta_1^{(-c)} \geq \log \rho_1^{-1}$.

Remark 1.3. The $\theta_i^{-\max}$ for $i = 1, 2$ are computed from their definitions as

$$\begin{aligned} \theta_1^{-\max} &= \log \frac{1}{2\lambda_1} \left(1 - 2\sqrt{\mu_2(\lambda_2 + \delta)} + \zeta_1^{(-)}\right), \\ \theta_2^{-\max} &= \log \frac{1 - 4(\lambda_1\mu_1 + (\lambda_2 + \delta)\mu_2) + \zeta_2^{(-)}}{8\lambda_1(\lambda_2 + \delta)}, \end{aligned}$$

where

$$\begin{aligned} \zeta_1^{(-)} &= \sqrt{1 + 4(\mu_2(\lambda_2 + \delta) - \sqrt{\mu_2(\lambda_2 + \delta)} - \lambda_1\mu_1)}, \\ \zeta_2^{(-)} &= \sqrt{(1 - 4(\lambda_1\mu_1 + (\lambda_2 + \delta)\mu_2))^2 - 64(\lambda_2 + \delta)\lambda_1\mu_1\mu_2}. \end{aligned}$$

Similarly, letting $z = e^{\theta_1}$ and $\xi = e^{\theta_2}$ in (1.27),

$$\lambda_2 z + \mu_1 z \xi^{-1} + \mu_2 z^{-1} + (\lambda_1 + \delta) z^{-1} \xi = 1, \quad (1.32)$$

$$\lambda_2 z + \mu_1 + \mu_2 z^{-1} + (\lambda_1 + \delta) z^{-1} \xi = 1. \quad (1.33)$$

Solving these equations for $z \neq 1$, we have $z = \xi = \rho_2^{-1}$. For $z = \rho_2^{-1}$, (1.32) yields $\xi = \rho_2^{-1}$, $\frac{\mu_1}{\lambda_1 + \delta} \rho_2^{-1}$. Reminding that

$$\theta_1^{+\max} = \max\{\log z; \text{(1.32) holds.}\}, \quad \theta_2^{+\max} = \max\{\log \xi; \text{(1.32) holds.}\},$$

we have that $\theta_1^{(+c)} \geq \log \rho_2^{-1}$ and

$$\left(\theta_1^{(+c)}, \theta_2^{(+c)}\right) = \begin{cases} (\log \rho_2^{-1}, \log \rho_2^{-1}), & \mu_1 > \lambda_1 + \delta \\ (\theta_1^{+\max}, \theta_2^{+\max}), & \mu_1 \leq \lambda_1 + \delta. \end{cases} \quad (1.34)$$

We also consider to solve (1.28). In this case, let $\xi = e^{\eta_2}$, $z_1 = e^{-\eta_1^{(-)}}$ and $z_2 = e^{\eta_1^{(+)}}$. Then, (1.28) becomes

$$\lambda_1 z_1 + \mu_2 z_1 \xi^{-1} + \mu_1 z_1^{-1} + (\lambda_2 + \delta) z_1^{-1} \xi = 1, \quad (1.35)$$

$$\lambda_2 z_2 + \mu_1 z_2 \xi^{-1} + \mu_2 z_2^{-1} + (\lambda_1 + \delta) z_2^{-1} \xi = 1, \quad (1.36)$$

$$\left(\lambda_1 + \frac{\delta}{2}\right) z_1 + \mu_2 z_1 \xi^{-1} + \mu_1 z_2 \xi^{-1} + \left(\lambda_2 + \frac{\delta}{2}\right) z_2 = 1. \quad (1.37)$$

These equations have been solved in [3]. That is, if $z \neq 1$, then $\xi = \rho^{-2}$ and $z_1 = z_2 = \rho^{-1}$. For $\xi = \rho^{-2}$, the first equation has solutions $z_1 = \rho^{-1}$, $\frac{\eta_1 + \delta}{\eta_2} \rho^{-1}$, and

the second equation yields $z_2 = \rho^{-1}, \frac{\gamma_2 + \delta}{\gamma_1} \rho^{-1}$. In this case, $\eta_2^{(c)}$ is obtained as the maximum ξ that satisfies (1.35), (1.36) and

$$\left(\lambda_1 + \frac{\delta}{2}\right) z_1 + \mu_2 z_1 \xi^{-1} + \mu_1 z_2 \xi^{-1} + \left(\lambda_2 + \frac{\delta}{2}\right) z_2 \leq 1. \quad (1.38)$$

Thus, we need to solve a convex optimization problem. We here already know that $(z_1, z_2, \xi) = (1, 1, 1), (\rho^{-1}, \rho^{-1}, \rho^{-2})$ are the extreme points of the constrains. To identify the latter point on the convex curves (1.35) and (1.36), it is convenient to introduce the following classifications:

$$\gamma_2 + \delta > \gamma_1, \quad \gamma_1 + \delta > \gamma_2, \quad (1.39)$$

$$\gamma_2 + \delta \leq \gamma_1, \quad \gamma_1 + \delta > \gamma_2, \quad (1.40)$$

$$\gamma_2 + \delta > \gamma_1, \quad \gamma_1 + \delta \leq \gamma_2, \quad (1.41)$$

where we exclude the case that $\gamma_2 + \delta \leq \gamma_1$ and $\gamma_1 + \delta \leq \gamma_2$, which is impossible since $\delta > 0$. Note that (1.39) is equivalent to

$$|\gamma_1 - \gamma_2| < \delta,$$

which is introduced and called strongly pooled in [6].

We now find $\eta_2^{(c)}$ by solving the convex optimization problem.

Lemma 1.8. If the strongly pooled condition (1.39) holds, then

$$\eta_2^{(c)} = \log \rho^{-2}, \quad \eta_1^{(-c)} = \eta_1^{(+c)} = \log \rho^{-1}.$$

Otherwise, if (1.40) holds, then

$$(\eta_2^{(c)}, \eta_1^{(-c)}, \eta_1^{(+c)}) = \left(\theta_2^{-\max}, \log \frac{e^{\eta_2^{(c)}}}{2(\lambda_1 e^{\eta_2^{(c)}} + \mu_2)}, \arg \max_{(\theta_1, \eta_2^{(c)}) \in \mathcal{D}_0^{(+)}} \theta_1 \right),$$

and, if (1.41) holds, then

$$(\eta_2^{(c)}, \eta_1^{(-c)}, \eta_1^{(+c)}) = \left(\theta_2^{+\max}, \arg \max_{(\theta_1, \eta_2^{(c)}) \in \mathcal{D}_0^{(-)}} \theta_1, \log \frac{e^{\eta_2^{(c)}}}{2(\lambda_2 e^{\eta_2^{(c)}} + \mu_1)} \right).$$

We defer the proof of this lemma to Appendix B.

We next consider to apply Theorem 1.3 to the generalized join shortest queue. To this end, we introduce another classifications.

$$\rho_1 < \rho, \quad \rho_2 < \rho, \quad (1.42)$$

$$\rho_1 \geq \rho, \quad \rho_2 < \rho, \quad (1.43)$$

$$\rho_1 < \rho, \quad \rho_2 \geq \rho, \quad (1.44)$$

where we do not consider the case that $\rho_1 \geq \rho$ and $\rho_2 \geq \rho$, which is impossible since $\delta > 0$. The condition (1.42) is referred to as a weakly pooled condition in [6].

Under the conditions (1.39) and (1.42), the asymptotic decay of

$$P(\min(Q_1, Q_2) = n, Q_1 - Q_2 = \ell), \quad n \rightarrow \infty$$

is shown to be exactly geometric with decay rate ρ^2 for each fixed ℓ in [3], [6]. This is the only known results for the decay rate for the minimum of the two queues. Using the two sets of the classifications, we can answer to the decay rate problem for all the cases but for the weak decay rates.

Theorem 1.5. For the generalized join shortest queue with two queues, suppose that the stability conditions $\rho < 1$, $\rho_1 < 1$ and $\rho_2 < 1$ are satisfied. Then, the weak decay rate r_2 exists for the minimum of the two queues in the sense of marginal distribution as well as jointly with each fixed difference of the two queues, and one of the following three cases occurs.

(g1) If (1.39) holds, then either one of the following cases happens.

(g1a) (1.42) implies $r_2 = \rho^2$.

(g1b) (1.43) implies $r_2 = \frac{\lambda_2 + \delta}{\mu_2} \rho_1$.

(g1c) (1.44) implies $r_2 = \frac{\lambda_1 + \delta}{\mu_1} \rho_2$.

(g2) If (1.40) holds, then either one of the following cases happens.

(g2a) (1.42) implies $r_2 = \begin{cases} e^{-\theta_2^- \max}, & \eta_1^{(+c)} \leq \theta_1^{(+c)} \\ \frac{\lambda_1 + \delta}{\mu_1} \rho_2, & \eta_1^{(+c)} > \theta_1^{(+c)}. \end{cases}$

(g2b) (1.43) implies

$$r_2 = \begin{cases} e^{-\theta_2^- \max}, & \eta_1^{(-c)} < \log \rho_1^{-1}, \eta_1^{(+c)} < \theta_1^{(+c)} \\ \frac{\lambda_2 + \delta}{\mu_2} \rho_1, & \eta_1^{(-c)} \geq \log \rho_1^{-1}, \eta_1^{(+c)} < \theta_1^{(+c)} \\ \frac{\lambda_1 + \delta}{\mu_1} \rho_2, & \eta_1^{(-c)} < \log \rho_1^{-1}, \eta_1^{(+c)} \geq \theta_1^{(+c)} \\ \min\left(\frac{\lambda_2 + \delta}{\mu_2} \rho_1, \frac{\lambda_1 + \delta}{\mu_1} \rho_2\right), & \eta_1^{(-c)} \geq \log \rho_1^{-1}, \eta_1^{(+c)} \geq \theta_1^{(+c)}. \end{cases}$$

(g2c) (1.44) implies $r_2 = \frac{\lambda_1 + \delta}{\mu_1} \rho_2$.

(g3) If (1.41) holds, then either one of the following cases happens.

(g3a) (1.42) implies $r_2 = \begin{cases} e^{-\theta_2^+ \max}, & \eta_1^{(-c)} \leq \theta_1^{(-c)} \\ \frac{\lambda_2 + \delta}{\mu_2} \rho_1, & \eta_1^{(-c)} > \theta_1^{(-c)}. \end{cases}$

$$(g3b) \text{ (1.43) implies } r_2 = \frac{\lambda_2 + \delta}{\mu_2} \rho_1.$$

(g3c) (1.44) implies

$$r_2 = \begin{cases} e^{-\theta_2^{+\max}}, & \eta_1^{(-c)} < \theta_1^{(-c)}, \eta_1^{(+c)} < \log \rho_2^{-1} \\ \frac{\lambda_2 + \delta}{\mu_2} \rho_1, & \eta_1^{(-c)} \geq \theta_1^{(-c)}, \eta_1^{(+c)} < \log \rho_2^{-1} \\ \frac{\lambda_1 + \delta}{\mu_1} \rho_2, & \eta_1^{(-c)} < \theta_1^{(-c)}, \eta_1^{(+c)} \geq \log \rho_2^{-1} \\ \min\left(\frac{\lambda_2 + \delta}{\mu_2} \rho_1, \frac{\lambda_1 + \delta}{\mu_1} \rho_2\right), & \eta_1^{(-c)} \geq \theta_1^{(-c)}, \eta_1^{(+c)} \geq \log \rho_2^{-1}. \end{cases}$$

Furthermore, the decay rates are exactly geometric for the cases (g1), (g2) unless $\eta_1^{(-c)} = \theta_1^{-\max}$ and (g3) unless $\eta_1^{(+c)} = \theta_1^{+\max}$.

Proof. This theorem is concluded applying Theorem 1.3 together with Lemma 1.7 for $(\theta_1^{(sc)}, \theta_2^{(sc)})$ for $s = \pm$ and Lemma 1.8. We first consider case (g1a). In this case, we suppose that the strongly pooled condition (1.39) and the weakly pooled condition (1.42) hold, then $\theta_1^{(sc)} \geq \eta_1^{(sc)}$ for $s = \pm$ from (1.31), (1.34) and Lemma 1.6. Hence, either one of C(1,1), C(1,2) or C(2,1) occurs in Lemma 1.7, which implies that $r_2 = \alpha_2^{-1} = e^{-\eta_2^{(c)}} = \rho^2$.

We next consider (g1b). In this case, (1.39) and (1.43) are assumed. Note that $\rho_1 \geq \rho$ in (1.43) implies that

$$\mu_2 \geq \frac{\mu_1}{\lambda_1} (\lambda_2 + \delta) > \lambda_2 + \delta.$$

Hence, we always have $\theta_1^{(-c)} = \log \rho_1^{-1}$ from (1.31) in this case. Since $\log \rho_1^{-1} \leq \log \rho^{-1} = \eta_1^{(-c)}$ and $\eta_1^{+c} = \log \rho^{-1} < \log \rho_2^{-1} \leq \theta_1^{(+c)}$, we have (g1b) from (c5) or (c8) of Lemma 1.7.

The other cases are similarly proved. So, we omit their details. \square

To visualize the results of Theorem 1.5, we draw equations (1.26) and (1.27) on the (θ_1, θ_2) plane simultaneously for some examples. We here consider the four cases (g1a), (g1b), (g2a) and (g2b).

These four cases are given in Figs. 1.6 and 1.7. In case (g1a) of Fig. 1.6,

$$\lambda_1 = \frac{1}{16}, \quad \lambda_2 = \frac{3}{16}, \quad \delta = \frac{1}{8}, \quad \mu_1 = \frac{1}{4}, \quad \mu_2 = \frac{3}{8},$$

which implies that $\rho_1 = \frac{1}{4}$, $\rho_2 = \frac{1}{2}$ and $\rho = \frac{3}{5}$. In case (g1b),

$$\lambda_1 = \frac{6}{29}, \quad \lambda_2 = \frac{4}{29}, \quad \delta = \frac{1}{29}, \quad \mu_1 = \frac{10}{29}, \quad \mu_2 = \frac{8}{29},$$

which implies that $\rho_1 = 0.6$, $\rho_2 = 0.5$ and $\rho = \frac{11}{18}$.

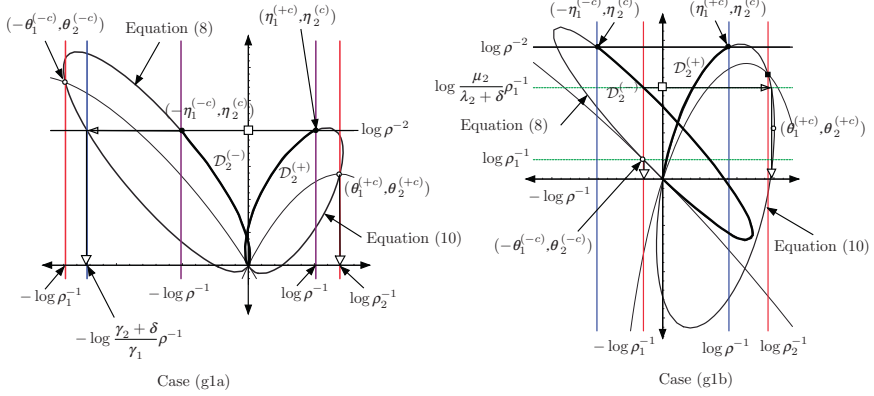


Fig. 1.6 The decay rates for strongly pooled (1.39): case (g1a) for (1.42) and case (g1b) for (1.43).

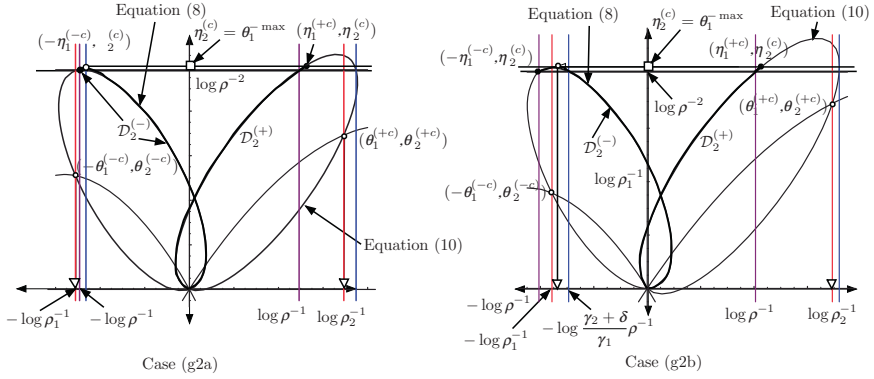


Fig. 1.7 The decay rates for not strongly pooled (1.40): case (g2a) for (1.42) and case (g2b) for (1.43).

Figure 1.7 shows the case where the weakly pooled condition (1.39) does not hold. In case (g2a), we set

$$\lambda_1 = \frac{9}{170}, \quad \lambda_2 = \frac{51}{170}, \quad \delta = \frac{1}{17}, \quad \mu_1 = \frac{1}{17}, \quad \mu_2 = \frac{9}{17},$$

which implies that $\rho_1 = 0.9$, $\rho_2 = \frac{17}{30}$ and $\rho = 0.7$. This example shows that the strongly pooled condition (1.39) does not imply the weakly pooled condition (1.42). In case (g2b),

$$\lambda_1 = \frac{7}{30}, \quad \lambda_2 = \frac{2}{15}, \quad \delta = \frac{1}{30}, \quad \mu_1 = \frac{10}{30}, \quad \mu_2 = \frac{8}{30},$$

which implies that $\rho_1 = 0.7$, $\rho_2 = 0.5$ and $\rho = \frac{2}{3}$.

Similarly to Theorem 1.5, we can get the following corollary for the decay rates for the difference of the two queues. We omit its proof since it is parallel to the arguments in Theorem 1.5.

Corollary 1.2. Under the assumptions of Theorem 1.5, the weak decay rates r_- and r_+ for the difference $Q_2 - Q_1$ in the negative and positive directions, respectively, exist in the sense of marginal distributions as well as jointly with each fixed minimum of the two queues, and we have the following cases, where $(\theta_1^{(-c)}, \theta_2^{(-c)})$ and $(\theta_1^{(+c)}, \theta_2^{(+c)})$ are given by (1.31) and (1.34), respectively, and

$$t_-(v) = \min\{z^{-1}; (1.29) \text{ for } \xi = v^{-1}\}, \quad t_+(v) = \min\{z^{-1}; (1.32) \text{ for } \xi = v^{-1}\}.$$

(h1) If (1.39) holds, then either one of the following cases happens.

(h1a) (1.42) implies

$$r_- = \begin{cases} e^{-\theta_1^{(-c)}}, & \theta_2^{(-c)} \leq \log \rho^{-2} \\ \frac{\gamma_2}{\gamma_1 + \delta} \rho, & \theta_2^{(-c)} > \log \rho^{-2}, \end{cases} \quad (1.45)$$

$$r_+ = \begin{cases} e^{-\theta_1^{(+c)}}, & \theta_2^{(+c)} \leq \log \rho^{-2} \\ \frac{\gamma_1}{\gamma_2 + \delta} \rho, & \theta_2^{(+c)} > \log \rho^{-2}. \end{cases} \quad (1.46)$$

(h1b) (1.43) implies with $r_2 = \frac{(\lambda_2 + \delta)}{\mu_2} \rho_1$ that

$$r_- = \rho_1, \quad r_+ = \begin{cases} e^{-\theta_1^{(+c)}}, & \theta_2^{(+c)} \leq \log r_2^{-1} \\ t_+(r_2), & \theta_2^{(+c)} > \log r_2^{-1}. \end{cases} \quad (1.47)$$

(h1c) If (1.44) implies with $r_2 = \frac{(\lambda_1 + \delta)}{\mu_1} \rho_2$ that

$$r_- = \begin{cases} e^{-\theta_1^{(-c)}}, & \theta_2^{(-c)} \leq \log r_2^{-1}, \\ t_-(r_2), & \theta_2^{(-c)} > \log r_2^{-1}, \end{cases} \quad r_+ = \rho_2. \quad (1.48)$$

(h2) If (1.40) holds, then either one of the following cases happens.

(h2a) (1.42) implies

$$(r_-, r_+) = \begin{cases} (e^{-\theta_1^{(-c)}}, \min(e^{-\theta_1^{(+c)}}, t_+(e^{-\eta_2^{(c)}}))), & \theta_1^{(+c)} \geq \eta_1^{(+c)} \\ (e^{-\theta_1^{(-c)}}, \rho_2), & \theta_1^{(+c)} < \eta_1^{(+c)}. \end{cases} \quad (1.49)$$

(h2b) (1.43) implies

$$(r_-, r_+) = \begin{cases} (\rho_1, e^{-\theta_1^{(+c)}}), & \eta_1^{(-c)} < \log \rho_1^{-1}, \eta_1^{(+c)} < \theta_1^{(+c)} \\ (\rho_1, \min(e^{-\theta_1^{(+c)}}, t_+(r_2))), & \eta_1^{(-c)} \geq \log \rho_1^{-1}, \eta_1^{(+c)} < \theta_1^{(+c)} \\ (\min(e^{-\theta_1^{(-c)}}, t_-(r_2)), \rho_2), & \eta_1^{(-c)} < \log \rho_1^{-1}, \eta_1^{(+c)} \geq \theta_1^{(+c)} \\ (\rho_1, \min(e^{-\theta_1^{(+c)}}, t_+(r_2))), & \eta_1^{(-c)} \geq \log \rho_1^{-1}, \eta_1^{(+c)} \geq \theta_1^{(+c)} \end{cases} \quad (1.50)$$

and $\frac{\lambda_2 + \delta}{\mu_2} \rho_1 < \frac{\lambda_1 + \delta}{\mu_1} \rho_2$

$$\begin{cases} (\min(e^{-\theta_1^{(-c)}}, t_-(r_2)), \rho_2), & \eta_1^{(-c)} \geq \log \rho_1^{-1}, \eta_1^{(+c)} \geq \theta_1^{(+c)} \\ \text{and } \frac{\lambda_2 + \delta}{\mu_2} \rho_1 \geq \frac{\lambda_1 + \delta}{\mu_1} \rho_2, \end{cases}$$

where $t_+ = \max\{z; (1.32) \text{ for } \xi = \log r_2^{-1}\}$ and $r_2 = \frac{(\lambda_2 + \delta)}{\mu_2} \rho_1$.

(h2c) (1.44) implies with $r_2 = \frac{\lambda_1 + \delta}{\mu_1} \rho_2$ that

$$(r_-, r_+) = \begin{cases} (e^{-\theta_1^{(-c)}}, \rho_2), & \theta_2^{(-c)} < \log r_2^{-1} \\ (t_-(r_2), \rho_2), & \theta_2^{(-c)} \geq \log r_2^{-1}. \end{cases} \quad (1.51)$$

Furthermore, the decay rates are exactly geometric unless either $r_- = e^{-\theta_1^{(-c)}}$ with $\theta_1^{(-c)} = \theta_1^{-\max}$ or $r_+ = e^{-\theta_1^{(+c)}}$ with $\theta_1^{(+c)} = \theta_1^{+\max}$.

Remark 1.4. In this corollary, the case that (1.41) holds is not considered. However, this case can be easily obtained by interchanging the roles of queues 1 and 2 in case (h2).

1.6 Remarks on Existence Results

We remark how our results include the existence results. The exactly geometric rate $r_2 = \rho^2$ is obtained under the conditions (1.39) and (1.42) in [3], [6]. Our results cover all the possible cases although the decay rates are generally of the weak sense. We also note that there are some errors in Theorem 3.2 of [3]. They can be corrected by Corollary 1.2. Namely, the additional conditions (3.16) and (3.18) there are not sufficient to get the decay rates. They are used for all the terms in the sums of (3.15) and (3.17) to be positive. However, this is different from the corresponding eigenvectors to be positive. The right conditions are $\theta_2^{(+c)} \geq \log \rho^{-2}$ and $\theta_2^{(-c)} \geq \log \rho^{-2}$, respectively, where $\theta_2^{(-c)}$ and $\theta_2^{(+c)}$ are given in (1.31) and (1.34), respectively.

1.7 Conclusions

In this paper, we completely characterized the weak tail decay rates in terms of the transition probabilities for the stationary distribution of the two sided $DQBD$ process (Theorems 1.3). For the exactly geometric decay, we find sufficient conditions, which are close to necessary conditions (Theorem 1.4). We then apply those results to the generalized join shortest queue with two waiting lines, whose decay rate problem has been only solved under some special conditions such as the weakly and strongly pooled conditions in the literature. We completely answer to this problem by finding the weak decay rates of the stationary distributions of the minimum of the two queues and their difference for all cases (Theorem 1.5 and Corollary 1.2). It is notable that the strongly and weakly pooled conditions still play the important role for finding the decay rate for the minimum of two queues. That is, the decay rate crucially changes according to whether or not those two conditions are satisfied.

Acknowledgments I am grateful to Yiqiang Zhao for his careful reading the original manuscript of this chapter and many invaluable comments. I also think Mr. Hiroyuki Yamakata for computing some numerical values. This research is supported in part by JSPS under grant No. 18510135.

Appendix 1

We prove Theorem 1.2. Let $\mathbf{x} = (\dots, x_{-1}, x_0, x_1, \dots)$ be the right positive invariant vector of $A_*^{(2)}(z)$. Then, we have

$$\begin{aligned}
 x_n &= p_{1*}^{(-)}(z)x_{n-1} + p_{0*}^{(-)}(z)x_n + p_{(-1)*}^{(-)}(z)x_{n+1}, & n \leq -2, \\
 x_{-1} &= p_{1*}^{(-)}(z)x_{-2} + p_{0*}^{(-)}(z)x_{-1} + p_{(-1)*}^{(2)}(z)x_0, \\
 x_0 &= p_{1*}^{(-)}(z)x_{-1} + p_{0*}^{(2)}(z)x_0 + p_{(-1)*}^{(+)}(z)x_1, & (1.52) \\
 x_1 &= p_{1*}^{(2)}(z)x_0 + p_{0*}^{(+)}(z)x_1 + p_{(-1)*}^{(+)}(z)x_2, \\
 x_n &= p_{1*}^{(+)}(z)x_{n-1} + p_{0*}^{(+)}(z)x_n + p_{(-1)*}^{(+)}(z)x_{n+1}, & n \geq 2.
 \end{aligned}$$

For $s = \pm$, let $w_1^{(s)}$ and $w_2^{(s)}$ be the solutions of the following quadratic equation:

$$p_{(-1)*}^{(s)}(z)w^2 - (1 - p_{0*}^{(s)}(z))w + p_{1*}^{(s)}(z) = 0. \quad (1.53)$$

Then \mathbf{x} must have the following forms:

$$x_n = \begin{cases} x_{-1}(w_1^{(-)})^{n+1} + (x_{-2} - x_{-1}(w_1^{(-)})^{-1}) \sum_{\ell=-n+2}^0 (w_1^{(-)})^{-\ell} (w_2^{(-)})^{n+2+\ell}, & n \leq -2 \\ x_1(w_1^{(+)})^{n-1} + (x_2 - x_1 w_1^{(+)}) \sum_{\ell=0}^{n-2} (w_1^{(+)})^{\ell} (w_2^{(+)})^{n-2-\ell}, & n \geq 2. \end{cases} \quad (1.54)$$

By the irreducibility assumption in (i), $p_{1*}^{(s)}(z) > 0$ and $p_{(-1)*}^{(s)}(z) > 0$. Furthermore, the positivity of $x_n, w_1^{(s)}, w_2^{(s)}$ must be real numbers. Hence, from the fact

$$w_1^{(s)} w_2^{(s)} = \frac{p_{1*}^{(s)}(z)}{p_{(-1)*}^{(s)}(z)} > 0, \quad (1.55)$$

$w_1^{(s)}$ and $w_2^{(s)}$ must be positive. This implies that \mathbf{x} is nonnegative if and only if

$$x_{-2} w_1^{(-)} \geq x_{-1}, \quad x_2 \geq x_1 w_1^{(+)}. \quad (1.56)$$

From (1.52), we have

$$x_{-2} = \frac{1}{p_{1*}^{(-)}(z)} \left(x_{-1} - p_{0*}^{(-)}(z) x_{-1} - p_{(-1)*}^{(2)}(z) x_0 \right), \quad (1.57)$$

$$x_2 = \frac{1}{p_{(-1)*}^{(+)}(z)} \left(x_1 - p_{1*}^{(2)}(z) x_0 - p_{0*}^{(+)}(z) x_1 \right). \quad (1.58)$$

Substituting these x_{-2} and x_2 into (1.56) yields

$$\begin{aligned} & \left((1 - p_{0*}^{(-)}(z)) w_1^{(-)} - p_{1*}^{(-)}(z) \right) x_{-1} - p_{(-1)*}^{(2)}(z) w_1^{(-)} x_0 \geq 0, \\ & \left((1 - p_{0*}^{(+)}(z)) - p_{(-1)*}^{(+)}(z) w_1^{(+)} \right) x_1 - p_{1*}^{(2)}(z) x_0 \geq 0. \end{aligned}$$

Since $w_1^{(s)}$ satisfies (1.53), we have

$$\begin{aligned} & p_{(-1)*}^{(-)}(z) w_1^{(-)} x_{-1} - p_{(-1)*}^{(2)}(z) x_0 \geq 0, \\ & p_{1*}^{(+)}(z) x_1 - p_{1*}^{(2)}(z) w_1^{(+)} x_0 \geq 0. \end{aligned}$$

Using (1.55), this is equivalent to

$$p_{1*}^{(-)}(z) x_{-1} - p_{(-1)*}^{(2)}(z) w_2^{(-)} x_0 \geq 0, \quad (1.59)$$

$$p_{(-1)*}^{(+)}(z) x_1 - p_{1*}^{(2)}(z) (w_2^{(+)})^{-1} x_0 \geq 0. \quad (1.60)$$

Hence, letting

$$\eta_2 = \log z, \quad \eta_1^{(s)} = -\log w_2^{(s)},$$

we have (1.14), (1.15) and (1.16).

We next show that these conditions are also sufficient. Suppose that there are $\eta_2 \geq 0$ and $\eta_1^{(s)}$ with $s = \pm 1$ satisfying (1.14), (1.15) and (1.16). Then, we can find $u^{(s)}$ with $s = \pm 1$ such that

$$u^{(-)} + \varphi_0^{(2)}(\eta_2) + u^{(+)} = 1, \quad u^{(-)} \geq \varphi_{-1}^{(2)}(\eta_2)e^{-\eta_1^{(-)}}, \quad u^{(+)} \geq \varphi_1^{(2)}(\eta_2)e^{\eta_1^{(+)}}.$$

Let $x_0 = 1$, and define x_{-1} and x_1 as $x_{-1} = \frac{u^{(-)}}{p_{1*}^{(-)}(z)}$, $x_1 = \frac{u^{(+)}}{p_{(-1)*}^{(+)}(z)}$. Hence, letting

$z = e^{\eta_2}$ and $w_2^{(s)} = e^{-\eta_1^{(s)}}$ with $s = \pm 1$, we have (1.59) and (1.60). Then, defining x_{-2} , x_2 and x_n by (1.57), (1.58) and (1.54), respectively, we revive (1.52). Hence, we indeed find the positive left eigenvector \mathbf{x} of $A_*^{(2)}(z)$. This proves the first part of the theorem. The remaining parts are obvious from (1.54) and Lemma 1.2. \square

Appendix 2

We prove Lemma 1.8. Define the following functions on \mathbb{R}_+^3 , where $\mathbb{R}_+ = (0, \infty)$,

$$\begin{aligned} f(z_1, z_2, \xi) &= \xi, \\ g_1(z_1, z_2, \xi) &= (\lambda_1 \xi + \mu_2)z_1^2 + \mu_1 \xi + (\lambda_2 + \delta)\xi^2 - z_1 \xi, \\ g_2(z_1, z_2, \xi) &= (\lambda_2 \xi + \mu_1)z_2^2 + \mu_2 \xi + (\lambda_1 + \delta)\xi^2 - z_2 \xi, \\ h(z_1, z_2, \xi) &= \left(\lambda_1 + \frac{\delta}{2}\right)z_1 \xi + \mu_2 z_1 + \mu_1 z_2 + \left(\lambda_2 + \frac{\delta}{2}\right)z_2 \xi - \xi. \end{aligned}$$

Obviously, all the functions are convex. Then, Lemma 1.8 is obtained by the following optimization problem. In particular, $\eta_2^{(c)}$ is obtained as the logarithm of the maximum value of f .

$$\text{miximize } f(z_1, z_2, \xi),$$

subject to

$$g_1(z_1, z_2, \xi) = 0, \quad g_2(z_1, z_2, \xi) = 0, \quad h(z_1, z_2, \xi) \leq 0, \quad (1.61)$$

$$z_1 > 0, \quad z_2 > 0, \quad \xi \geq 1. \quad (1.62)$$

This is a convex optimization problem, and (1.61) is satisfied with equality only if $(z_1, z_2, \xi) = (1, 1, 1)$ or $(\rho^{-1}, \rho^{-1}, \rho^{-2})$ (see Lemma 3.2 of [3]). By D , we denote the set of all feasible solutions satisfying the constraints (1.61) and (1.62). Clearly, D is closed and bounded in \mathbb{R}_+^3 . For convenience, let

$$D_i = \{z_i; (z_1, z_2, \xi) \in D\}, \quad i = 1, 2.$$

Since $\{(z_i, \xi) \in \mathbb{R}_+^2; g_i(z_1, z_2, \xi) \leq 0\}$ is a convex set, $g_i(z_1, z_2, \xi) = 0$ have two solutions counting multiplicity for each ξ and each $i = 1, 2$ if the solution exists. Hence, there exist at most four points $(z_1, z_2, \xi) \in D$ for each ξ .

We show that D is a connected curve with end points $(1, 1, 1)$ and $(\rho^{-1}, \rho^{-1}, \rho^{-2})$ if D has three points at least. Suppose that this is not true. Let $(z_1^\circ, z_2^\circ, \xi^\circ) \in D$ be the third point other than the above end points. Then, we must have $h(z_1^\circ, z_2^\circ, \xi^\circ) < 0$. This implies that the point (z_1°, z_2°) is in the interior of the set

$$\{(z_1, z_2) \in \mathbb{R}_+^2; h(z_1, z_2, \xi) \leq 0\},$$

for $\xi = \xi^\circ$, which is a polyhedral for each ξ and its region is continuously increased as ξ is increased. Hence, there exists a connected curve which passes through $(z_1^\circ, z_2^\circ, \xi^\circ)$ as an inner point. This curve must have $(1, 1, 1)$ and $(\rho^{-1}, \rho^{-1}, \rho^{-2})$ as its end points since otherwise we arrive at the contradiction that there is a point other than those points such that $h = 0$ holds.

Let us consider the cases for (1.39) and (1.40) separately. Here, we do not consider the case for (1.41) since it is symmetric to the case for (1.40). Denote the solutions of $g_i(z_1, z_2, \xi) = 0$ for each ξ by $\underline{z}_i(\xi)$ and $\bar{z}_i(\xi)$, where $\underline{z}_i(\xi) \leq \bar{z}_i(\xi)$. First, assume that (1.39) holds. Then $(\underline{z}_1(\rho^{-2}), \underline{z}_2(\rho^{-2}), \rho^{-2}) = (\rho^1, \rho^{-1}, \rho^{-2}) \in D$ and $\bar{z}_i(\rho^{-2}) \notin D_i$ for $i = 1, 2$. Hence, f is maximized at $(\rho^{-1}, \rho^{-1}, \rho^{-2})$. We next assume (1.40). Then, we have $(\bar{z}_1(\rho^{-2}), \bar{z}_2(\rho^{-2}), \rho^{-2}) = (\rho^{-1}, \rho^{-1}, \rho^{-2}) \in D$ and $(\underline{z}_1(\rho^{-2}), \underline{z}_2(\rho^{-2}), \rho^{-2}) \in D$ since $\underline{z}_1(\rho^{-2}) \leq \bar{z}_1(\rho^{-2})$. If $\underline{z}_1(\rho^{-2}) = \bar{z}_1(\rho^{-2})$, we can reduce the problem to the case for (1.39). Otherwise, D has three points at least, so it is a connected curve with end points $(1, 1, 1)$ and $(\rho^{-1}, \rho^{-1}, \rho^{-2})$ as shown above. This concludes that f is maximized at $(\underline{z}_1(\xi^*), \underline{z}_2(\xi^*), \xi^*)$ such that $\underline{z}_1(\xi^*) = \bar{z}_1(\xi^*)$. Since ξ^* must be the maximum value of ξ satisfying $g_1(z_1, z_2, \xi) = 0$, $\eta_2^{(c)} = \theta_2^{-\max}$. This completes the proof. \square

It may be notable that we can also solve the optimization problem by applying Karush-Kuhn-Tucker necessary conditions (e.g., see Sect. 4.3.7 of [11]). However, the present solution is more informative since the feasible region D is identified to be a connected curve.

References

1. M. Miyazawa, "Tail decay rates in a doubly QBD process," Submitted for publication.
2. M. Miyazawa, "Doubly QBD process and a solution to the tail decay rate problem," in *Proc. the Second Asia-Pacific Symposium on Queueing Theory and Network Applications*, pp. 33-42, 2007.
3. H. Li, M. Miyazawa, and Y. Q. Zhao, "Geometric decay in a QBD process with countable background states with applications to shortest queues," *Stochastic Models*, vol. 23, pp. 413-438, 2007.
4. G. Latouche and V. Ramaswami, *Introduction to Matrix Analytic Methods in Stochastic Modeling*. Philadelphia: American Statistical Association and the Society for Industrial and Applied Mathematics, 1999.

5. M. F. Neuts, *Matrix-Geometric Solutions in Stochastic Models*. Baltimore: Johns Hopkins University Press, 1981.
6. R. D. Foley and D. R. McDonald, "Join the shortest queue: stability and exact asymptotics," *The Annals of Applied Probability*, vol. 11, pp. 569-607, 2001.
7. A. A. Puhalskii and A. A. Vladimirov, "A large deviation principle for join the shortest queue," *Mathematics of Operations Research*, vol. 32, pp. 700-710, 2007.
8. M. Miyazawa and Y. Q. Zhao, "The stationary tail asymptotics in the GI/G/1 type queue with countably many background states," *Advances in Applied Probability*, vol. 36, pp. 1231-1251, 2004.
9. E. Seneta, *Nonnegative Matrices and Markov Chains*, Second Edition. New York: Springer-Verlag, 1981.
10. R. D. Foley and D. R. McDonald, "Bridges and networks: Exact asymptotics," *The Annals of Applied Probability*, vol. 15, pp. 542-586, 2005.
11. M. S. Bazaraa, H. D. Sherali, and C. M. Shetty, *Nonlinear Programming: Theory and Algorithms*. New York: Wiley, 1993.