Heavy-traffic analysis of k-limited polling systems

Boon, M.A.A.; Winands, E.M.M.

Published: 01/01/2013

Document Version
Publisher's PDF, also known as Version of Record (includes final page, issue and volume numbers)

Please check the document version of this publication:

- A submitted manuscript is the author's version of the article upon submission and before peer-review. There can be important differences between the submitted version and the official published version of record. People interested in the research are advised to contact the author for the final version of the publication, or visit the DOI to the publisher's website.
- The final author version and the galley proof are versions of the publication after peer review.
- The final published version features the final layout of the paper including the volume, issue and page numbers.

Link to publication

Citation for published version (APA):

General rights
Copyright and moral rights for the publications made accessible in the public portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

- Users may download and print one copy of any publication from the public portal for the purpose of private study or research.
- You may not further distribute the material or use it for any profit-making activity or commercial gain
- You may freely distribute the URL identifying the publication in the public portal?

Take down policy
If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

Download date: 01. Dec. 2018
Heavy-traffic analysis of $k$-limited polling systems

M. Boon, E. Winands
ISSN 1389-2355
Heavy-traffic analysis of $k$-limited polling systems

M.A.A. Boon*  E.M.M. Winands †
marko@win.tue.nl  e.m.m.winands@uva.nl

January 22, 2013

Abstract

In this paper we study a two-queue polling model with zero switch-over times and $k$-limited service (serve at most $k_i$ customers during one visit period to queue $i$, $i = 1, 2$) in each queue. The arrival processes at the two queues are Poisson, and the service times are exponentially distributed. By increasing the arrival intensities until one of the queues becomes critically loaded, we derive exact heavy-traffic limits for the joint queue-length distribution using a singular-perturbation technique. It turns out that the number of customers in the stable queue has the same distribution as the number of customers in a vacation system with Erlang-$k_2$ distributed vacations. The queue-length distribution of the critically loaded queue, after applying an appropriate scaling, is exponentially distributed. Finally, we show that the two queue-length processes are independent in heavy traffic.

Keywords: polling model, queue lengths, heavy traffic, perturbation

Mathematics Subject Classification: 60K25, 90B22

1 Introduction

This paper considers a two-queue $k$-limited polling model with exponentially distributed service times and zero switch-over times. Under the $k$-limited strategy the server continues working until either a predefined number of $k_i$ customers is served at queue $i$ or until the queue becomes empty, whichever occurs first. The interest for this model is fueled by a number of applications in the fields of communication and logistics (see, e.g., [4, 7, 24]). In the present paper, we consider the heavy-traffic scenario, in which one of the queues becomes critically loaded with the other queue remaining stable.

Although the number of papers on polling systems is impressive, hardly any exact results for polling systems with the $k$-limited service policy have been obtained. This can be explained by the fact that the $k$-limited strategy does not satisfy a well-known branching property for polling systems, independently discovered by Fuhrmann [12] and Resing [23]. Groenendijk [13] and Ibe [14] give an explicit Laplace-Stieltjes Transform for the waiting-time distribution in a two-queue 1-limited/exhaustive system. For two-queue systems where both queues are served according to the 1-limited discipline, the

*Eurandom and Department of Mathematics and Computer Science, Eindhoven University of Technology, P.O. Box 513, 5600MB Eindhoven, The Netherlands
†University of Amsterdam, Korteweg-de Vries Institute for Mathematics, Science Park 904, 1098 XH Amsterdam, The Netherlands
problem of finding the queue length distribution can be shown to translate into a boundary value problem [5, 6, 8, 10]. For general $k$, an exact evaluation for the queue-length distribution is known in two-queue exhaustive/$k$-limited systems (see [18, 21, 22, 24]). The paper of Lee [18] for the two-queue exhaustive/$k$-limited system without setup times is also the only paper analysing the heavy-traffic behaviour for the $k$-limited discipline. More specifically, he studies the limiting regime where the exhaustive queue remains stable, while the $k$-limited queue becomes critically loaded in the limit.

The main contribution of the present paper is that we derive heavy-traffic asymptotics for $k$-limited polling models. That is, by increasing the arrival intensities until one of the queues becomes critically loaded, we derive exact heavy-traffic limits for the joint queue-length distribution in a two-queue $k$-limited polling model using a singular-perturbation technique. In this way, we derive the lowest-order asymptotic to the joint queue-length distribution in terms of a small positive parameter measuring the closeness of the system to instability. See Knessl and Tier [16] for an excellent survey of applications of the perturbation technique to queueing models. It is noteworthy that our paper is inspired by the manner in which Morrison and Borst [20] apply this technique to a model with interacting queues.

Furthermore, the results obtained in the present paper provide new insights into the heavy-traffic behaviour of $k$-limited polling systems. It is shown that the number of customers in the stable queue has the same distribution as the number of customers in a vacation system with Erlang-$k$ distributed vacations, while the scaled queue-length distribution of the critically loaded queue is exponentially distributed. Finally, we prove that the two queue-length processes are independent in heavy traffic. These results do not only generalise those derived in [18] for the special case of exhaustive/$k$-limited service, but are also obtained via a fundamentally different singular-perturbation approach. Finally, the singular-perturbation technique can also be extended to an $N$-queue system ($N \geq 2$) with one queue becoming critically loaded. In this limiting regime the stable queues have the same joint distribution of a $k$-limited polling model with $N-1$ queues and an extended switch-over time.

The paper is structured as follows. In the next section we introduce the model and notation. In Section 3 we apply a perturbation technique to study the system under heavy-traffic conditions and derive the limiting scaled joint queue-length distribution. In Section 4 we interpret the results and give some suggestions on further research. The appendices contain some lengthy derivations required for the analysis in Section 3.

## 2 Model description and notation

We consider a polling model consisting of two queues, $Q_1$ and $Q_2$, that are alternately visited by a single server. Throughout this paper, the subscript $i$ will always be used to refer to one of the queues, meaning that it always takes on the values 1 or 2. When a server arrives at $Q_i$, it serves at most $k_i$ customers. When $k_i$ customers have been served or $Q_i$ becomes empty, whichever occurs first, the server switches to the other queue. We assume that switching from one queue to the other requires no time. If the other queue turns out to be empty, the server switches back and serves, again, at most $k_i$ customers. If both queues are empty, the server waits until the first arrival and switches to the corresponding queue (say, $Q_j$) to start another visit period of at most $k_j$ customers, $j = 1, 2$. Customers arrive at $Q_i$ according to a Poisson process with intensity $\lambda_i$. We assume that the service times of customers in $Q_i$ are independent and exponentially distributed with parameter $\mu_i$. We denote the load of the system by $\rho = \rho_1 + \rho_2$, where $\rho_i = \lambda_i / \mu_i$. For a polling model without switch-over
times, the stability condition is \( \rho < 1 \) [9, 11]. Furthermore, we assume that

\[ \frac{\lambda_1}{k_1} < \frac{\lambda_2}{k_2}. \]  

(2.1)

This assumption is discussed in more detail in the next section.

The number of customers in \( Q_i \) at time \( t, t \geq 0 \), is denoted by \( N_i(t) \). In order to describe the queue length process as a Markov process, we use the approach of Blanc [3], introducing a supplementary variable \( H(t) \) which takes on values 1, 2, ..., \( k_1 + k_2 \). The variable \( H(t) \) is used to determine the server position (\( Q_1 \) or \( Q_2 \)) at time \( t \) and the number of customers that can be served before the server has to switch to the next queue. When \( 1 \leq H(t) \leq k_1 \), this means that the server is serving \( Q_1 \) at time \( t \), and that the customer in service is the \( H(t) \)-th customer being served during the present visit period. If \( k_1 + 1 \leq H(t) \leq k_1 + k_2 \), the server is serving the \((H(t) - k_1)\)-th customer in \( Q_2 \). Now, \((N_1(t), N_2(t), H(t))\) is a Markov process. Assuming that \( \rho < 1 \), define the stationary probabilities

\[ p(n_1, n_2, h) := \lim_{t \to \infty} P(N_1(t) = n_1, N_2(t) = n_2, H(t) = h), \]

and define the steady-state queue lengths \( N_i \).

3 Analysis

We study the heavy-traffic limit of the joint queue-length process \((N_1, N_2)\) by increasing the arrival rate \( \lambda_2 \), while keeping \( \lambda_1 \) fixed. When \( \rho \) tends to 1, Assumption (2.1) implies that \( Q_2 \) will become critically loaded, whereas \( Q_1 \) remains stable due to the fact that at most \( k_2 \) customers are served during each visit period at \( Q_2 \). In case \( \lambda_1 / k_1 = \lambda_2 / k_2 \) in the limit, both queues would become critically loaded simultaneously and the system behaviour is different from the limiting behaviour found in the present paper. In fact, the limiting queue-length behaviour for that specific case remains an open problem. We discuss this topic briefly in Section 4.

We use a single-perturbation method to find the queue-length distributions in heavy-traffic \((\rho \uparrow 1)\). First, we write down the balance equations of our model and apply a perturbation to the arrival rate of \( Q_2 \) to these equations, in the case that this queue is close to becoming critically loaded. By solving the system of balance equations for the lowest order terms, we find the queue length distribution of the stable queue, \( Q_1 \). By solving for the first-order and second-order terms, we also obtain a differential equation for the scaled number of customers in \( Q_2 \), which we can solve to show that this number converges to an exponential distribution.

3.1 Balance equations

The balance equations for a polling model with exponentially distributed service times and \( k \)-limited service at each of the queues are given by Blanc [3]. For completeness, we present these equations for our two-queue model below.

3
(\lambda_1 + \lambda_2 + \mu_2) p(0, n_2, k_1 + 1) = \lambda_2 p(0, n_2 - 1, k_1 + 1) + \mu_2 p(0, n_2 + 1, k_1 + k_2) \\
+ \sum_{h=1}^{k_1} \mu_1 p(1, n_2, h), \quad (3.1a)

(\lambda_1 + \lambda_2 + \mu_2) p(0, n_2, h_2) = \lambda_2 p(0, n_2 - 1, h_2) + \mu_2 p(0, n_2 + 1, h_2 - 1), \quad (3.1b)

(\lambda_1 + \lambda_2 + \mu_1) p(1, n_2, 1) = \lambda_2 p(1, n_2 - 1, 1) + \mu_2 p(1, n_2 + 1, k_1 + k_2), \quad (3.1c)

(\lambda_1 + \lambda_2 + \mu_1) p(1, n_2, h_1) = \lambda_2 p(1, n_2 - 1, h_1) + \mu_1 p(2, n_2, h_1 - 1), \quad (3.1d)

(\lambda_1 + \lambda_2 + \mu_1) p(n_1 + 1, n_2, 1) = \lambda_1 p(n_1, n_2, 1) + \lambda_2 p(n_1 + 1, n_2 - 1, 1) + \mu_2 p(n_1 + 1, n_2 + 1, k_1 + k_2), \quad (3.1e)

(\lambda_1 + \lambda_2 + \mu_1) p(n_1 + 1, n_2, h_1) = \lambda_1 p(n_1 + 1, n_2, h_1) + \lambda_2 p(n_1 + 1, n_2 - 1, h_1) + \mu_1 p(n_1 + 2, n_2, h_1 - 1), \quad (3.1f)

(\lambda_1 + \lambda_2 + \mu_2) p(n_1, n_2, k_1 + 1) = \lambda_1 p(n_1 - 1, n_2, k_1 + 1) + \lambda_2 p(n_1, n_2 - 1, k_1 + 1) + \mu_1 p(n_1 + 1, n_2, k_1), \quad (3.1g)

(\lambda_1 + \lambda_2 + \mu_2) p(n_1, n_2, h_2) = \lambda_1 p(n_1 - 1, n_2, h_2) + \lambda_2 p(n_1, n_2 - 1, h_2) + \mu_2 p(n_1, n_2 + 1, h_2 - 1), \quad (3.1h)

for \(n_1 = 1, 2, \ldots; n_2 = 2, 3, \ldots; h_1 = 2, 3, \ldots, k_1,\) and \(h_2 = k_1 + 2, \ldots, k_1 + k_2.\) Note that (3.1a)-(3.1h) are not all balance equations. We have omitted all equations for \(n_2 = 0\) and \(n_2 = 1,\) since it will turn out that these do not play a role after the perturbation. The intuitive explanation is that \(N_2(t)\) will tend to infinity as \(Q_2\) becomes critically loaded and the probabilities \(p(n_1, n_2, h)\) become negligible for low values of \(n_2.\)

### 3.2 Perturbation

From the stability condition we have that the system becomes unstable as \(\lambda_2/\mu_2 \uparrow 1 - \lambda_1/\mu_1,\) which means that the arrival rate \(\lambda_2\) approaches \(\mu_2 (1 - \lambda_1/\mu_1).\) Therefore we will assume that

\[
\lambda_2 = \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) - \delta \omega, \quad \omega > 0, 0 < \delta \ll 1.
\]

At the end of this section we will take an appropriate choice for the constant \(\omega,\) which will influence the limit of the scaled queue length in \(Q_2.\)

Let \(\xi = \delta n_2,\) and

\[
p(n_1, \xi/\delta, h) = \delta \phi_{n_1, h}(\xi, \delta), \quad 0 < \xi = O(1), h = 1, 2, \ldots, k_1 + k_2. \quad (3.3)
\]

The next step is to substitute (3.2) and (3.3) in the balance equations (3.1a)-(3.1h), and take the Taylor series expansion with respect to \(\delta.\) For reasons of compactness, we only show the intermediate results for Equation (3.1h) as an illustration:

\[
(\lambda_1 + \mu_2) \phi_{n_1, h_2}(\xi, \delta) - \lambda_1 \phi_{n_1 - 1, h_2}(\xi, \delta) - \mu_2 \phi_{n_1, h_2 - 1}(\xi + \delta, \delta) = \\
\left(\mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) - \delta \omega\right) (\phi_{n_1, h_2}(\xi - \delta, \delta) - \phi_{n_1, h_2}(\xi, \delta)).
\]
Taking the Taylor series yields:

\[
(\lambda_1 + \mu_2)\phi_{n_1,h_2}(\xi, \delta) - \lambda_1 \phi_{n_1-1,h_2}(\xi, \delta) - \mu_2 \left( \phi_{n_1-1,h_2}(\xi, \delta) + \delta \frac{\partial \phi_{n_1,h_2-1}(\xi, \delta)}{\partial \xi} + \frac{\delta^2}{2} \frac{\partial^2 \phi_{n_1,h_2-1}(\xi, \delta)}{\partial \xi^2} \right) - \left( \mu_2(1 - \frac{\lambda_1}{\mu_1}) - \delta \omega \right) \left( \delta \frac{\partial \phi_{n_1,h_2}(\xi, \delta)}{\partial \xi} - \frac{\delta^2}{2} \frac{\partial^2 \phi_{n_1,h_2}(\xi, \delta)}{\partial \xi^2} \right) + O(\delta^3). \tag{3.4}
\]

Note that \( \lambda_2 \) (or \( \mu_2(1 - \frac{\lambda_1}{\mu_1}) - \delta \omega \) after the substitution) only plays a role in this equation for \( O(\delta) \) terms and higher. It is readily verified that this is the case for all balance equations. We now expand in powers of \( \delta \), and let

\[
\phi_{n_1,h}(\xi, \delta) = \phi_{n_1,h}^{(0)}(\xi) + \delta \phi_{n_1,h}^{(1)}(\xi) + O(\delta^2). \tag{3.5}
\]

We also define the corresponding generating functions

\[ \tilde{Q}_h(z, \xi, \delta) := \sum_{n_1=0}^{\infty} \phi_{n_1,h}(\xi, \delta) z^{n_1}, \quad \tilde{Q}_h^{(j)}(z, \xi) := \sum_{n_1=0}^{\infty} \phi_{n_1,h}^{(j)}(\xi) z^{n_1}, \quad j = 0, 1, 2, \ldots. \]

In the next subsections we first equate the lowest order terms of the resulting equations to find an expression for (the generating function of) \( \phi_{n_1,h}^{(0)}(\xi) \), and subsequently we equate the first-order and second-order terms to find the scaled queue-length distribution of \( Q_2 \).

### 3.3 Equating the lowest-order terms

Equating the lowest-order terms of the balance equations, after substituting (3.2), (3.3), and (3.5), results in the following equations.

\[
(\lambda_1 + \mu_2)\phi_{0,k_1+1}^{(0)}(\xi) = \mu_2\phi_{0,k_1+k_2}^{(0)}(\xi) + \sum_{h=1}^{k_1} \mu_1 \phi_{1,h}^{(0)}(\xi), \tag{3.6a}
\]

\[
(\lambda_1 + \mu_2)\phi_{0,h_2}^{(0)}(\xi) = \mu_2\phi_{0,h_2-1}^{(0)}(\xi), \tag{3.6b}
\]

\[
(\lambda_1 + \mu_1)\phi_{1,1}^{(0)}(\xi) = \mu_2\phi_{1,k_1+k_2}^{(0)}(\xi), \tag{3.6c}
\]

\[
(\lambda_1 + \mu_1)\phi_{1,h_1}^{(0)}(\xi) = \mu_1 \phi_{2,h_1-1}^{(0)}(\xi), \tag{3.6d}
\]

\[
(\lambda_1 + \mu_1)\phi_{1,k_1+1}^{(0)}(\xi) = \lambda_1 \phi_{n_1-1,k_1+1}^{(0)}(\xi) + \mu_2 \phi_{n_1+1,k_1+k_2}^{(0)}(\xi), \tag{3.6e}
\]

\[
(\lambda_1 + \mu_1)\phi_{1,n_1+1,h_1}^{(0)}(\xi) = \lambda_1 \phi_{n_1-1,h_1}^{(0)}(\xi) + \mu_1 \phi_{n_1+2,h_1-1}^{(0)}(\xi), \tag{3.6f}
\]

\[
(\lambda_1 + \mu_2)\phi_{n_1-1,k_1+1}^{(0)}(\xi) = \lambda_1 \phi_{n_1-1,k_1+1}^{(0)}(\xi) + \mu_1 \phi_{n_1+1,k_1}^{(0)}(\xi), \tag{3.6g}
\]

\[
(\lambda_1 + \mu_2)\phi_{n_1-1,h_2}^{(0)}(\xi) = \lambda_1 \phi_{n_1-1,h_2}^{(0)}(\xi) + \mu_2 \phi_{n_1,h_2-1}^{(0)}(\xi), \tag{3.6h}
\]

for \( n_1 = 1, 2, \ldots; h_1 = 2, 3, \ldots, k_1 \), and \( h_2 = k_1 + 2, \ldots, k_1 + k_2 \).

Note that \( \sum_{n_1=0}^{\infty} \sum_{h=1}^{k_1+k_2} \phi_{n_1,h}^{(0)}(\xi) \neq 1 \). For this reason we introduce \( P_0(\xi) \) and \( \pi_{n_1,h}^{(0)} \), with

\[
\phi_{n_1,h}^{(0)}(\xi) = \pi_{n_1,h}^{(0)} P_0(\xi), \quad \text{and} \quad \sum_{n_1=0}^{\infty} \sum_{h=1}^{k_1+k_2} \pi_{n_1,h}^{(0)} = 1, \tag{3.7}
\]
for \( n_1 = 0, 1, 2, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2 \). Careful inspection of these balance equations reveals that equations (3.6a)-(3.6h) describe the behaviour of a single-server vacation queue with the following properties:

P1. the arrival process is Poisson with intensity \( \lambda_1 \),

P2. the service times are exponentially distributed with mean \( 1/\mu_1 \),

P3. the service discipline is \( k \)-limited service with service limit \( k_1 \),

P4. the vacations are Erlang-\( k_2 \) distributed with parameter \( \mu_2 \),

P5. whenever the server finds the system empty upon return from a vacation, it immediately starts another vacation.

This system has been studied in the literature (cf. [19]) and, in general, no closed-form expressions for the steady-state queue-length probabilities can be obtained. However, it is possible to find the probability generating function (PGF) of the queue-length distribution. Define

\[
\tilde{L}^{(0)}(z) := \sum_{h=1}^{k_1+k_2} \tilde{L}_h^{(0)}(z), \quad \text{where} \quad \tilde{L}_h^{(0)}(z) := \sum_{n=0}^{\infty} \pi_{n,h}^{(0)} z^n, \quad h = 1, \ldots, k_1 + k_2,
\]

and

\[
\tilde{G}(z) := \frac{\mu_1}{\lambda_1(1-z) + \mu_1}, \quad \tilde{H}(z) := \frac{\mu_2}{\lambda_1(1-z) + \mu_2}.
\]

It is easily seen that \( \tilde{G}(z) \) and \( \tilde{H}(z) \) are the PGFs of the number of arrivals during respectively one service, and during one stage of the vacation (which consists of \( k_2 \) exponential stages). It follows from (3.6a)-(3.6h) that

\[
\tilde{L}^{(0)}_{k_1+k_2}(z) = \tilde{H}(z)^{k_2} \left[ \frac{\pi_{0,k_1+k_2}^{(0)} (1 - (\tilde{G}(z)/z)^{k_1}) + \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} \pi_{1,h}^{(0)} (1 - (\tilde{G}(z)/z)^{k_1-h})}{1 - (\tilde{G}(z)/z)^{k_1} \tilde{H}(z)^{k_2}} \right],
\]

(3.9)

\[
\tilde{L}^{(0)}_{h_1}(z) = \frac{z \mu_2}{\mu_1} \left( \tilde{G}(z)/z \right)^{h_1} \tilde{H}(z)^{k_1} \left[ \frac{\pi_{0,k_1+k_2}^{(0)} (\tilde{L}^{(0)}_{k_1+k_2}(z) - \pi_{0,k_1+k_2}^{(0)}) + \sum_{h=1}^{h_1-1} \pi_{1,h}^{(0)} z (\tilde{G}(z)/z)^{h_1-h}}{1 - (\tilde{G}(z)/z) \tilde{H}(z)^{k_1}} \right],
\]

(3.10)

\[
\tilde{L}^{(0)}_{h_2}(z) = \tilde{H}(z)^{h_2-k_1} \left[ \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} \pi_{1,h}^{(0)} + \pi_{0,k_1+k_2}^{(0)} + \frac{\mu_1}{\mu_2} \tilde{L}^{(0)}_{k_1}(z) \right],
\]

(3.11)

for \( h_1 = 1, \ldots, k_1 \) and \( h_2 = k_1 + 1, \ldots, k_1 + k_2 - 1 \). A derivation of (3.9)-(3.11) can be found in Appendix A. These equations still contain \( k_1 \) unknowns: \( \pi_{1,1}^{(0)}, \pi_{1,2}^{(0)}, \ldots, \pi_{1,k_1-1}^{(0)}, \) and \( \pi_{0,k_1+k_2}^{(0)} \). See Appendix A for more details on how to eliminate them using Rouché’s Theorem. Foregoing the derivation of the limiting behaviour of \( Q_2 \) we already would like to mention that these unknowns do not play a role therein.

We conclude from equating the lowest-order terms of the balance equations (3.1a)-(3.1h), after substituting (3.2), (3.3), and (3.5), that

\[
\sum_{n_1=0}^{\infty} \sum_{h=1}^{k_1+k_2} \phi_{n_1,h}^{(0)}(\xi) z^{n_1} = \tilde{L}^{(0)}(z) P_0(\xi),
\]

(3.12)

where \( P_0(\xi) \) still has to be determined. Consequently, in heavy traffic, the queue length of the stable queue (\( Q_1 \)) has the same distribution as the queue length in a vacation system with Erlang(\( k_2 \)).
distributed vacations with parameter \( \mu_2 \), exponential service times with parameter \( \mu_1 \), and \( k_1 \)-limited service. The assumption that we have Poisson arrivals and first-come-first-served service implies that we can use the distributional form of Little’s law to obtain (the Laplace-Stieltjes transform of) the waiting-time distribution of customers in \( Q_1 \) (see, for example, Keilson and Servi [15]).

**Remark 3.1** In this paper we assume that \( \lambda_1/k_1 < \lambda_2/k_2 \), causing \( Q_2 \) to become critically loaded when \( \lambda_2 \) is being increased. We have implicitly used this assumption when solving the balance equations (3.1a)-(3.1h). It is well-known that the vacation system described by these equations is stable if and only if

\[
\lambda_1 \mathbb{E}[C] < k_1,
\]

where \( \mathbb{E}[C] \) is the mean cycle time, i.e., the mean length of one visit period plus one vacation. Denoting the length of a vacation by \( S \), we have \( \mathbb{E}[C] = \mathbb{E}[S]/(1 - \rho_1) = k_2/(\mu_2(1 - \lambda_1/\mu_1)) \). When substituting this in (3.13), we indeed obtain exactly the same inequality as (2.1) after substituting (3.2) and letting \( \delta \downarrow 0 \).

**Remark 3.2** Another interesting observation is that one could consider more general ways of varying the arrival rates in order to let \( Q_2 \) become critically loaded. To this end, we introduce \( \lambda_1^* \) and \( \lambda_2^* \) such that

\[
\frac{\lambda_1^*}{k_1} < \frac{1}{\mu_1 + \frac{\lambda_2^*}{\mu_2}}, \quad \text{or equivalently} \quad \frac{\lambda_2^*}{k_2} > \frac{1}{\mu_1 + \frac{\lambda_1^*}{\mu_2}}.
\]

(3.14)

We now let \( \lambda_1 \to \lambda_1^* \) and \( \lambda_2 \to \lambda_2^* \) for \( \delta \downarrow 0 \), with

\[
\frac{\lambda_1}{\mu_1} + \frac{\lambda_2}{\mu_2} = 1 - \delta \omega^*, \quad \omega^* > 0, \ 0 < \delta \ll 1.
\]

(3.15)

Any arbitrary way in which we let \( \lambda_1 \) and \( \lambda_2 \) approach respectively \( \lambda_1^* \) and \( \lambda_2^* \), for \( \delta \downarrow 0 \), will cause \( Q_2 \) to become critically loaded (because of assumption (3.14)). All results obtained in this paper will still be valid, by choosing \( \omega^* = \omega/\mu_2 \).

### 3.4 Equating the first-order terms

In this section we study, and solve, the system of equations that results from equating the first-order terms of the perturbed balance equations. For notational reasons, we define

\[
\psi_{n_1,h}^{(1)}(\xi) := \phi_{n_1,h}^{(1)}(\xi) + \phi_{n_1,h}^{(0)}(\xi), \quad \text{where} \quad \phi_{n_1,h}^{(0)}(\xi) := \frac{d\phi_{n_1,h}^{(0)}(\xi)}{d\xi},
\]

for \( n_1 = 0, 1, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2 \). The resulting set of equations for the probabilities \( \phi_{n_1,h}(\xi) \) is given below.
\( (\lambda_1 + \mu_2)\phi_{0,k_1+1}^{(1)}(\xi) = \mu_2\psi_{0,k_1+k_2}^{(1)}(\xi) + \sum_{h=1}^{k_1} \mu_1 \phi_{1,h}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{0,k_1+1}^{(0)}(\xi), \tag{3.16a} \)

\( (\lambda_1 + \mu_2)\phi_{0,h_2}^{(1)}(\xi) = \mu_2\psi_{0,h_2-1}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{0,h_2}^{(0)}(\xi), \tag{3.16b} \)

\( (\lambda_1 + \mu_1)\phi_{1,1}^{(1)}(\xi) = \mu_2\psi_{1,k_1+k_2}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{1,1}^{(0)}(\xi), \tag{3.16c} \)

\( (\lambda_1 + \mu_1)\phi_{1,h_1}^{(1)}(\xi) = \mu_1\phi_{2,h_1-1}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{1,h_1}^{(0)}(\xi), \tag{3.16d} \)

\( (\lambda_1 + \mu_1)\phi_{n_1+1,1}^{(1)}(\xi) = \lambda_1\phi_{n_1,1}^{(1)}(\xi) + \mu_2\psi_{n_1+1,k_1+k_2}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{n_1+1,1}^{(0)}(\xi), \tag{3.16e} \)

\( (\lambda_1 + \mu_1)\phi_{n_1+1,h_1}^{(1)}(\xi) = \lambda_1\phi_{n_1,h_1}^{(1)}(\xi) + \mu_1\phi_{n_2+1,h_1+1}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{n_1+1,h_1}^{(0)}(\xi), \tag{3.16f} \)

\( (\lambda_1 + \mu_2)\phi_{n_1,k_1+1}^{(1)}(\xi) = \lambda_1\phi_{n_1-1,k_1+1}^{(1)}(\xi) + \mu_1\phi_{n_1+1,k_1}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{n_1+1,k_1}^{(0)}(\xi), \tag{3.16g} \)

\( (\lambda_1 + \mu_2)\phi_{n_1,h_2}^{(1)}(\xi) = \lambda_1\phi_{n_1-1,h_2}^{(1)}(\xi) + \mu_2\psi_{n_2,h_2-1}^{(1)}(\xi) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \phi_{n_1+1,h_2}^{(0)}(\xi), \tag{3.16h} \)

for \( n_1 = 1, 2, \ldots; h_1 = 2, 3, \ldots, k_1, \) and \( h_2 = k_1 + 2, \ldots, k_1 + k_2. \) The solution to this system of equations, in terms of generating functions, can be found in Appendix B. This solution is used to derive the following relation

\[
\frac{\lambda_1}{\mu_1} \sum_{n=0}^{\infty} \sum_{h=1}^{k_1+k_2} \phi_{n,h}^{(1)}(\xi) - \sum_{n=0}^{\infty} \sum_{h_1=1}^{k_1} \phi_{n,h_1}^{(1)}(\xi) = \frac{\lambda_1 \mu_2}{\mu_1^2} P_0^{(1)}(\xi), \tag{3.17} \]

which turns out to play a key role in determining the HT limit of the joint queue-length distribution (see the next section).

### 3.5 Equating the second-order terms

In order to find an expression for \( P_0^{(1)}(\xi) \) and, consequently, solve (3.12), we consider the Taylor series of all perturbed balance equations. In this section we show that, fortunately, we only need to consider the sum of all these equations (such as Equation (3.4)) over all \( n_1 = 0, 1, 2, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2, \) and we consecutively consider the \( \Theta(1), \Theta(\delta), \) and \( \Theta(\delta^2) \) terms. Using the results we have obtained so far, we prove that:

1. all \( \Theta(1) \) terms cancel immediately,
2. the \( \Theta(\delta) \) terms cancel after expanding \( \phi_{n_1,h}(\xi, \delta) \) in powers of \( \delta \) (i.e., substituting (3.5)),
3. the equation that results from equating the \( \Theta(\delta^2) \) terms, can be solved to find \( P_0^{(1)}(\xi) \).

The above three results are proven in Propositions 3.3, 3.4, and 3.5.
Proposition 3.3 After taking the summation over all \( n_1 = 0, 1, 2, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2 \) of the Taylor series of all perturbed balance equations, the \( \mathcal{O}(1) \) terms cancel.

Proof:
We can follow the generating function approach, used in Section 3.3, but replace the probabilities \( \pi_{n_1}^{(0)} \), by \( \phi_{n_1,h}(\xi, \delta) \). This results in the same set of equations as (3.9)-(3.11), but with terms \( \tilde{Q}_h(z, \xi, \delta) \) and \( \phi_{n_1,h}(\xi, \delta) \) instead of \( \tilde{L}_h^{(0)}(z) \) and \( \pi_{n_1}^{(0)} \). Substituting \( z = 1 \) yields:

\[
\mu_2 \tilde{Q}_{h_2}(1, \xi, \delta) = \mu_2 \tilde{Q}_{h_2-1}(1, \xi, \delta), \tag{3.18}
\]

\[
\mu_2 \tilde{Q}_{k_1+1}(1, \xi, \delta) = \mu_1 \tilde{Q}_{k_1}(1, \xi, \delta) + \mu_1 \sum_{h=1}^{k_1-1} \phi_{1,h}(\xi, \delta) + \mu_2 \phi_{0,k_1+k_2}(\xi, \delta), \tag{3.19}
\]

\[
\mu_1 \tilde{Q}_{h_1}(1, \xi, \delta) = \mu_1 \tilde{Q}_{h_1-1}(1, \xi, \delta) - \mu_1 \phi_{1,k_1-1}(\xi, \delta), \tag{3.20}
\]

\[
\mu_1 \phi_{1,k_1+k_2}(1, \xi, \delta) = \mu_2 \phi_{0,k_1+k_2}(\xi, \delta), \tag{3.21}
\]

for \( h_1 = 2, \ldots, k_1 \) and \( h_2 = k_1 + 2, \ldots, k_1 + k_2 \). The summation of (3.18)-(3.21) over all \( h = 1, 2, \ldots, k_1 + k_2 \) cancels all terms.

Proposition 3.4 After taking the summation over all \( n_1 = 0, 1, 2, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2 \) of the Taylor series of all perturbed balance equations, substituting (3.5) and using the results from Appendix A, the \( \mathcal{O}(\delta) \) terms cancel.

Proof:
Define

\[
\phi'_{n_1,h}(\xi, \delta) := \frac{\partial \phi_{n_1,h}(\xi, \delta)}{\partial \xi},
\]

for \( n_1 = 0, 1, \ldots \) and \( h = 1, 2, \ldots, k_1 + k_2 \). Given the fact that the \( \mathcal{O}(1) \) terms cancel, taking the \( \mathcal{O}(\delta) \) terms leads to the following equations:

\[
0 = \mu_2 \phi'_{0,k_1+k_2}(\xi, \delta) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{0,k_1+1}(\xi, \delta),
\]

\[
0 = \mu_2 \phi'_{0,h_2-1}(\xi, \delta) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{0,h_2}(\xi, \delta),
\]

\[
0 = \mu_2 \phi'_{1,k_1+k_2}(\xi, \delta) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{1,1}(\xi, \delta),
\]

\[
0 = -\mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{1,h_1}(\xi, \delta),
\]

\[
0 = \mu_2 \phi'_{n_1+1,k_1+k_2}(\xi, \delta) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{n_1+1,1}(\xi, \delta),
\]

\[
0 = -\mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{n_1+1,1}(\xi, \delta),
\]

\[
0 = -\mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{n_1,k_1+1}(\xi, \delta),
\]

\[
0 = \mu_2 \phi'_{n_1,h_2-1}(\xi, \delta) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \phi'_{n_1,h_2}(\xi, \delta),
\]
for $h_1 = 2, \ldots, k_1$ and $h_2 = k_1 + 2, \ldots, k_1 + k_2$. Taking the generating functions of these equations and substituting $z = 1$ results in the following set of equations:

\begin{align*}
0 &= \mu_2 \Phi_{0,k_1+k_2}(\xi, \delta) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \tilde{Q}_{k_1+1}(1, \xi, \delta), \\
0 &= \mu_2 \tilde{Q}_{h_2-1}(1, \xi, \delta) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \tilde{Q}_{h_2}(1, \xi, \delta), \\
0 &= \mu_2 \left(\tilde{Q}_{k_1+k_2}(1, \xi, \delta) - \Phi_{0,k_1+k_2}(\xi, \delta)\right) - \mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \tilde{Q}_1(1, \xi, \delta), \\
0 &= -\mu_2 \left(1 - \frac{\lambda_1}{\mu_1}\right) \tilde{Q}_{h_1}(1, \xi, \delta),
\end{align*}

for $h_1 = 2, \ldots, k_1$ and $h_2 = k_1 + 2, \ldots, k_1 + k_2$. The PGF $\tilde{Q}_h(z, \xi, \delta)$ is the derivative of $\tilde{Q}_h(z, \xi, \delta)$ with respect to $\xi$.

The summation of (3.22)-(3.25) over all $h = 1, 2, \ldots, k_1 + k_2$ yields:

\begin{equation}
\mu_2 \sum_{h=1}^{k_1} \tilde{Q}_{h_1}(1, \xi, \delta) = \mu_2 \frac{\lambda_1}{\mu_1} \sum_{h=1}^{k_1+k_2} \tilde{Q}_h(1, \xi, \delta). 
\end{equation}

Apparenty, the $O(\delta)$ terms do not cancel (yet). However, after substituting (3.5), taking the $O(\delta)$ terms, and using (3.7), we obtain:

\begin{equation}
\mu_2 \sum_{h=1}^{k_1} \tilde{L}_{h_1}^{(0)}(1) P_0'(\xi) = \mu_2 \frac{\lambda_1}{\mu_1} \sum_{h=1}^{k_1+k_2} \tilde{L}_h^{(0)}(1) P_0'(\xi). 
\end{equation}

Since the (at this moment still unknown) terms $P_0'(\xi)$ cancel out, and since $\sum_{h=1}^{k_1+k_2} \tilde{L}_h^{(0)}(1) = 1$, Equation (3.27) reduces to

\[\sum_{h_1=1}^{k_1} \tilde{L}_{h_1}^{(0)}(1) = \rho_1,\]

which is indeed true (see (A.5)).

**Proposition 3.5** Taking the summation over all $n_1 = 0, 1, 2, \ldots$ and $h = 1, 2, \ldots, k_1 + k_2$ of the Taylor series of all perturbed balance equations and equating the $O(\delta^2)$ terms, yields the following differential equation for $P_0(\xi)$:

\begin{equation}
\omega P_0''(\xi) = -\left(\mu_2 + \frac{\lambda_1\mu_2(\mu_2 - \mu_1)}{\mu_1^2}\right) P_0''(\xi). 
\end{equation}

**Proof:**

Define

\[\phi''_{n_1,h}(\xi, \delta) := \frac{\partial^2 \phi_{n_1,h}(\xi, \delta)}{\partial \xi^2},\]

for $n_1 = 0, 1, \ldots$ and $h = 1, 2, \ldots, k_1 + k_2$. As before, the $O(1)$ terms cancel. From the proof of the Proposition 3.4 we have learned to include $O(\delta)$ terms as well, because multiplied by $\delta \phi_{n_1,h}'(\xi)$ these
the following three types of terms: 

\[ 0 = \mu_2 \left[ \phi_{0,l_1+k_2}^{(1)}(\xi) + \frac{1}{2} \phi_{0,l_1+k_2}^{(0)}(\xi) \right] - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{0,l_1+1}^{(1)}(\xi) - \frac{1}{2} \phi_{0,l_1+1}^{(0)}(\xi) \right] \\
+ \omega \phi_{0,l_1+1}^{(0)}(\xi) \delta, \]

\[ 0 = \mu_2 \left[ \phi_{0,h_2-1}^{(1)}(\xi) + \frac{1}{2} \phi_{0,h_2-1}^{(0)}(\xi) \right] - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{0,h_1}^{(1)}(\xi) - \frac{1}{2} \phi_{0,h_1}^{(0)}(\xi) \right] + \omega \phi_{0,h_1}^{(0)}(\xi), \]

\[ 0 = \mu_2 \left[ \phi_{1,k_1+k_2}^{(1)}(\xi) + \frac{1}{2} \phi_{1,k_1+k_2}^{(0)}(\xi) \right] - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{1,l_1+1}^{(1)}(\xi) - \frac{1}{2} \phi_{1,l_1+1}^{(0)}(\xi) \right] + \omega \phi_{1,l_1+1}^{(0)}(\xi), \]

\[ 0 = \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{n_1,1}^{(1)}(\xi) - \frac{1}{2} \phi_{n_1,1}^{(0)}(\xi) \right] + \omega \phi_{n_1,1}^{(0)}(\xi), \]

\[ 0 = \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{n_1+1,k_1+1}^{(1)}(\xi) - \frac{1}{2} \phi_{n_1+1,k_1+1}^{(0)}(\xi) \right] + \omega \phi_{n_1+1,k_1+1}^{(0)}(\xi), \]

\[ 0 = \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{n_1,1}^{(1)}(\xi) - \frac{1}{2} \phi_{n_1,1}^{(0)}(\xi) \right] + \omega \phi_{n_1,1}^{(0)}(\xi), \]

\[ 0 = \mu_2 \left[ \phi_{n_1,h_2-1}^{(1)}(\xi) + \frac{1}{2} \phi_{n_1,h_2-1}^{(0)}(\xi) \right] - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ \phi_{n_1,h_2-1}^{(1)}(\xi) - \frac{1}{2} \phi_{n_1,h_2-1}^{(0)}(\xi) \right] + \omega \phi_{n_1,h_2-1}^{(0)}(\xi), \]

for \( h_1 = 2, \ldots, k_1 \) and \( h_2 = k_1 + 2, \ldots, k_1 + k_2 \). As we have done already a couple of times before, we can use the generating functions to easily sum all of these equations. Each equation contains the following three types of terms: \( \phi_{n_1,b}^{(1)}(\xi) \), \( \phi_{n_1,b}^{(0)}(\xi) \), and \( \phi_{n_1,b}^{(0)}(\xi) \). We denote the corresponding generating functions with \( \tilde{Q}_{h_1}^{(0)}(z, \xi) \), \( \tilde{Q}_{h_1}^{(1)}(z, \xi) \), and \( \tilde{Q}_{h_1}^{(0)}(z, \xi) \). After summing all equations, we obtain the following equation:

\[ \mu_2 \sum_{h=1}^{k_1+k_2} \tilde{Q}_{h}^{(1)}(1, \xi) - \mu_2 \sum_{h=1}^{k_1} \tilde{Q}_{h}^{(1)}(1, \xi) + \frac{1}{2} \mu_2 \sum_{h=1}^{k_1} \tilde{Q}_{h}^{(0)}(1, \xi) + \frac{1}{2} \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \sum_{h=1}^{k_1+k_2} \tilde{Q}_{h}^{(0)}(1, \xi) + \omega \sum_{h=1}^{k_1+k_2} \tilde{Q}_{h}^{(0)}(1, \xi) = 0. \]

Note that the derivation of (3.29a) and (3.29b) follows closely the manner in which (3.26) has been derived. The last term (3.29c) follows directly from collecting the \( \omega \)-terms.

Using the results obtained in appendices the equation can be rewritten to

\[ \frac{\lambda_1}{\mu_1} \sum_{h=1}^{k_1} P_{0}^{(1)}(\xi) + \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) P_{0}^{(0)}(\xi) + \omega P_{0}^{(0)}(\xi) = 0. \]
The proof that (3.29a) can be rewritten to the first term in (3.30) can be found in Appendix B. The second and third term follow from (A.5) and the fact that \( \tilde{Q}_h^{(0)}(1, \xi) = \tilde{L}_h^{(0)}(1) P_0(\xi) \). Equation (3.30) can be rewritten to (3.28), which concludes this proof.

3.6 The scaled number of customers in the critically loaded queue

Now we can finally present the density of the scaled number of customers in \( Q_2 \), denoted by \( P_0(\xi) \). It is obtained by solving the differential equation (3.28):

\[
P_0(\xi) = \eta e^{-\eta \xi},
\]

with

\[
\frac{\omega}{\eta} = \mu_2 + \frac{\lambda_1 \mu_2 (\mu_2 - \mu_1)}{\mu_1^2}.
\]

We have used that \( \int_0^\infty P_0(\xi) d\xi = 1 \) and that \( \sum_{n_1=0}^{\infty} \sum_{n_2=0}^{\infty} \sum_{h=1}^{k_1+k_2} p(n_1, n_2, h) = 1 \). Without loss of generality, we may take \( \omega = \mu_2 \), which means that

\[
\frac{1}{\eta} = 1 - \frac{\lambda_1}{\mu_1} + \frac{\mu_2 \lambda_1}{\mu_1^2}.
\]

We motivate this choice for \( \omega \) by noting that we consider the scaled queue length \( \delta N_2 \) for \( \delta \downarrow 0 \). By choosing \( \omega = \mu_2 \), and using (3.2), our scaling becomes equivalent to considering the scaled queue length \( (1 - \rho) N_2 \), which is commonly used. Finally, by applying the multiclass distributional law of Bertsimas and Mourtzinou [2] it directly follows that the scaled waiting time at \( Q_2 \) also follows an exponential distribution with parameter \( \lambda_2 \eta \).

3.7 Main result

The analysis of the present section has the following immediate consequence for the joint (scaled) queue-length distribution in heavy traffic, which is the main result of this paper.

Main result: For \( \lambda_1/k_1 < \lambda_2/k_2 \) and \( \lambda_2 = \mu_2 ((1 - \rho_1) - \delta) \), we have:

\[
\lim_{\delta \downarrow 0} \mathbb{P}[N_1 \leq n_1, \delta N_2 \leq \xi] = \mathcal{L}(n_1) \left( 1 - e^{-\eta \xi} \right),
\]

where \( \mathcal{L}(\cdot) \) is the cumulative probability distribution of the queue length of a queueing system with multiple vacations satisfying Properties P1–P5, and \( \eta \) is given by (3.32).

4 Final remarks and suggestions for further research

Interpretation. The main result (3.33) derived in the preceding section has the following intuitively appealing interpretation:

1. The number of customers in the stable queue has the same distribution as the number of customers in a \( k \)-limited vacation system with Erlang-\( k_2 \) distributed vacations.
2. The scaled number of customers in the critically loaded queue is exponentially distributed with parameter $\eta$.

3. The number of customers in the stable queue and the (scaled) number of customers in the critically loaded queue are independent.

Below we explain these properties heuristically.

Property 1 can be explained by the fact that if $Q_2$ is in heavy-traffic, then exactly $k_2$ customers are served at this queue during each cycle. If we place an outside observer at $Q_1$, then, from his perspective, this queue behaves like a $k$-limited vacation model in heavy-traffic, where the vacation distribution is given by the convolutions of $k_2$ exponentially distributed service times distributions in $Q_2$.

For Property 2, we note that the total workload in the system equals the amount of work in an $M/G/1$ queue in which the two customer classes are combined into one customer class with arrival rate $\lambda_1 + \lambda_2$ and hyperexponentially distributed service times, i.e., the service time is with probability $\lambda_i/\left(\lambda_1 + \lambda_2\right)$ exponentially distributed with parameter $\mu_i$, $i = 1, 2$. Based on standard heavy-traffic results for the $M/G/1$ queue, this implies that the distribution of the scaled total workload converges to an exponential distribution with mean $\rho E[R]$, where $R$ is a residual service time. For a hyperexponential distribution, we have

$$E[R] = \frac{\lambda_1/\mu_1^2 + \lambda_2/\mu_2^2}{\rho},$$

which implies that the total (scaled) asymptotic workload is exponentially distributed with parameter $\lambda_1/\mu_1^2 + \lambda_2/\mu_2^2$. In heavy traffic, almost all customers are located in $Q_2$ so the total number of customers at this queue is exponentially distributed with mean $\mu_2\left(\lambda_1/\mu_1 + \lambda_2/\mu_2\right)$. Since $\lambda_2 \uparrow \mu_2\left(1 - \lambda_1/\mu_1\right)$, the scaled number of customers in $Q_2$ is exponentially distributed with parameter $\eta$. Using the multiclass distributional law of Bertsimas and Mourtzinou [2], it can be shown that the scaled asymptotic waiting time of customers in $Q_2$ is exponentially distributed with parameter $\lambda_2\eta$.

Finally, Property 3 follows from the time-scale separation in heavy-traffic which implies that the dynamics of the stable queue evolve at a much faster time scale than the dynamics of the critically loaded queue. Since the amount of “memory” of the stable queue asymptotically vanishes compared to that of the critically loaded queue, the number of customers in $Q_1$ and the scaled queue length of $Q_2$ become independent in the limit.

Two critically loaded queues. In the current paper we have analysed the heavy-traffic behaviour in case only $Q_2$ becomes critically loaded, i.e., when Assumption (2.1) is satisfied. The limiting regime in which both queues become saturated simultaneously ($\lambda_1/k_1 = \lambda_2/k_2$), shows fundamentally different system behaviour. That is, for general $\rho$ the waiting time at $Q_1$ is an (unknown) function of the visit time at $Q_2$ and $1/(1 - \rho)$. This implies that $\Theta(1 - \rho)$ variations in the visit time at $Q_2$ are relevant for the heavy-traffic behaviour at $Q_1$. More colloquially, it is not sufficient anymore to use a scaling that implies that exactly $k_2$ customers are served at $Q_2$ during each cycle, i.e., the probability that there are served less than $k_2$ customers cannot be neglected, when analysing the asymptotic behaviour of $Q_1$.

Further research. The analysis in this paper allows different kind of extensions. Firstly, one could consider phase-type interarrival-time or service-time distributions. The approach introduced in the
present paper may be extended, without complicating the analysis, to such a system. Another exten-
sion could be the introduction of switch-over times whenever the server switches between queues. Such an extension requires more severe adaptations to the approach and the analysis, and is the topic of a forthcoming paper. Finally, we want to mention that the singular-perturbation technique can also be applied to derive the HT analysis of a system consisting of more than two (say \(N\)) queues, with one queue becoming critically loaded. Following the lines of the current paper, one can show that in this limiting regime the stable queues have the same joint queue-length distribution as in a polling model with \(N - 1\) queues and an extended switch-over time, whereas the scaled queue-length distribution of the critically loaded queue is again exponentially distributed. As such, the results of the present paper provide a theoretical basis for the transformation of large polling systems into smaller systems for approximation purposes, cf. LaPadula and Levy [17].

Acknowledgement

The authors would like to thank Sem Borst and Onno Boxma for interesting discussions and valuable comments on earlier drafts of this paper.

Appendix

A A vacation model with \(k\)-limited service

In this appendix we study a queueing model with multiple vacations and \(k\)-limited service. The main goal is to find the PGF of the queue-length distribution, as to prove (3.9)-(3.11). At the end of this appendix some additional properties of this queue-length distribution are given, which will be used in Section 3.5 and in Appendix B.

The service times in this vacation model are exponentially distributed with parameter \(\mu_1\), and the vacation length is Erlang(\(k_2\)) distributed with parameter \(\mu_2\). The service discipline is \(k\)-limited service, with at most \(k_1\) customers being served during one visit period. Although the queue-length distribution for the case with generally distributed service and vacation times has been studied by Lee [19], we provide the proof here to keep the paper self-contained, but also because our state space is slightly different and we do not look at embedded epochs, yielding slightly different expressions than in [19].

The starting point is to obtain generating functions from the balance equations (3.6a)-(3.6h). Multiplying Equation (3.6h) with \(z^{n_1}\), summing over all \(n_1 = 1, 2, \ldots\), and adding Equation (3.6a), yields the following equation:

\[
\tilde{L}_{h_2}^{(0)}(z) = \tilde{H}(z)\tilde{L}_{h_2-1}^{(0)}(z),
\]

(A.1)

for \(h_2 = k_1 + 2, \ldots, k_1 + k_2\), where \(\tilde{H}(z)\) is defined in (3.8). The interpretation of (A.1) is that the number of customers in the system during a certain vacation stage is simply the number of customers present at the previous stage of the vacation, plus the arrivals during one (exponentially distributed) stage. Obviously, no customers leave the system during a vacation.

Multiplying Equation (3.6g) with \(z^{n_1}\), summing over all \(n_1 = 1, 2, \ldots\), and adding Equation (3.6a),
Conversely, the fraction of time that the system is serving customers is involving the roots and the numerator of (3.9) to eliminate these $k$ implies that the numerator of (3.9) should have these same roots [1]. Hence, we have a set of equations roots on and inside the unit circle. The requirement that (3.9) should be regular inside the unit circle, multiplied by $z$ yields the following equation:

$$\tilde{L}_{k+1}^{(0)}(z) = \tilde{H}(z) \left( \frac{\mu_1}{\mu_2} \sum_{h=1}^{k+1} \pi_{1,h}^{(0)} \, \pi_{0,k+2}^{(0)} + \frac{\mu_1}{\mu_2} \tilde{L}_k^{(0)}(z) \right). \quad \text{(A.2)}$$

Multiplying Equation (3.6f) with $z^{n_1+1}$, summing over all $n_1 = 1, 2, \ldots$, and adding Equation (3.6d) multiplied by $z$, yields the following equation:

$$\tilde{L}_{h_1}^{(0)}(z) = \frac{\tilde{G}(z)}{z} \left( \tilde{L}_{h-1}^{(0)}(z) - \pi_{1,h-1}^{(0)} \right), \quad \text{(A.3)}$$

for $h = 2, 3, \ldots, k_1$, where $\tilde{G}(z)$ is defined in (3.8).

Multiplying Equation (3.6e) with $z^{n_1+1}$, summing over all $n_1 = 1, 2, \ldots$, and adding Equation (3.6c) multiplied by $z$, yields the following equation:

$$\tilde{L}_1^{(0)}(z) = \frac{\mu_2}{\mu_1} \left( \tilde{L}_{k_1+k_2}^{(0)}(z) - \pi_{0,k_1+k_2}^{(0)} \right). \quad \text{(A.4)}$$

We now have $k_1 + k_2$ equations, each of which expresses $\tilde{L}_h^{(0)}(z)$ in terms of $\tilde{L}_{h-1}^{(0)}(z)$ (and $\tilde{L}_1^{(0)}(z)$ in terms of $\tilde{L}_{k_1+k_2}^{(0)}(z))$. Finally, we can solve these equations to find the expressions for $\tilde{L}_h^{(0)}(z)$, for $h = 1, 2, \ldots, k_1 + k_2$. The results are given in (3.9)-(3.11).

Note that there are still $k_1$ unknowns: $\pi_{1,h_1}^{(0)}$, for $h_1 = 1, \ldots, k_1 - 1$, and $\pi_{0,k_1+k_2}^{(0)}$. These can be found using the roots of the denominator of (3.9). Rouche’s Theorem states that the denominator has $k_1$ roots on and inside the unit circle. The requirement that (3.9) should be regular inside the unit circle, implies that the numerator of (3.9) should have these same roots [1]. Hence, we have a set of equations involving the roots and the numerator of (3.9) to eliminate these $k_1$ unknowns.

**Some additional properties.** In this paragraph we derive some results that are used throughout this paper, particularly in Section 3.5 and in Appendix B. From a balancing argument, we know that the fraction of time that the system is in a vacation is $1 - \rho$ (where, in this system, $\rho = \lambda_1/\mu_1$). Conversely, the fraction of time that the system is serving customers is $\rho$. Hence,

$$\sum_{h=1}^{k_1} \tilde{L}_h^{(0)}(1) = \frac{\lambda_1}{\mu_1}, \quad \sum_{h=k_1+1}^{k_1+k_2} \tilde{L}_h^{(0)}(1) = k_2 \tilde{L}_{k_1+k_2}^{(0)}(1) = 1 - \frac{\lambda_1}{\mu_1}. \quad \text{(A.5)}$$

Moreover, from (A.1) we know that $\tilde{L}_{h_2}^{(0)}(1) = \tilde{L}_{h_2-1}^{(0)}(1)$ for $h_2 = k_1 + 2, \ldots, k_1 + k_2$. It follows that $\tilde{L}_{h_2}^{(0)}(1) = \frac{1}{k_2} \left( 1 - \frac{\lambda_1}{\mu_1} \right), \quad h_2 = k_1 + 1, \ldots, k_1 + k_2$.

From (A.3) and (A.4) we now have

$$\tilde{L}_{h_1}^{(0)}(1) = \frac{\mu_2}{\mu_1} \left( \frac{1}{k_2} \left( 1 - \frac{\lambda_1}{\mu_1} \right) - \pi_{0,k_1+k_2}^{(0)} \right) - \sum_{i=1}^{k_1-1} \pi_{1,i}^{(0)}, \quad h_1 = 1, \ldots, k_1. \quad \text{(A.6)}$$
The following relation for the unknowns \( \pi_{0,k_1+k_2}^{(0)} \) and \( \pi_{1,k_1}^{(0)} \ (h_1 = 1, \ldots, k_1 - 1) \) can be derived by combining all of these results:

\[
k_1 \pi_{0,k_1+k_2}^{(0)} + \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} (k_1 - h) \pi_{1,h}^{(0)} = \frac{1}{k_2} \left[ k_1 \left( 1 - \frac{\lambda_1}{\mu_1} \right) - k_2 \frac{\lambda_1}{\mu_2} \right].
\]  

(A.7)

This relation turns out to be crucial to derive many results in this paper, without having to know the exact expressions for all of the individual probabilities.

**Remark A.1** A balancing argument has been the starting point to derive all of the above properties. A rigorous way to derive these results, is by using L'Hôpital’s rule on (3.9) to determine \( \tilde{L}_h^{(0)}(1) \), and subsequently deriving an expression for \( \sum_{h=1}^{k_1+k_2} \tilde{L}_h^{(0)}(1) \), which we know is equal to one.

**B The second-order perturbed balance equations**

The main goal of this appendix is to prove that (3.29a) can be written as the first term in (3.30). If we rearrange the summations slightly, we can write the equation that we need to prove as follows:

\[
\frac{\lambda_1}{\mu_1} \sum_{h_2=k_1+1}^{k_1+k_2} \tilde{Q}_h^{(1)}(1, \xi) - \mu_2 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \sum_{h_1=1}^{k_1} \tilde{Q}_h^{(1)}(1, \xi) = \frac{\lambda_1 \mu_2}{\mu_1^2} \phi_{n,0}''(\xi).
\]

(B.1)

This equation should follow from Equations (3.16a)-(3.16h). In order to prove it, we take the following steps:

1. First, we take the generating functions of Equations (3.16a)-(3.16h) to develop relations for \( \tilde{Q}_h^{(1)}(z, \xi) \) \( (h = 1, \ldots, k_1 + k_2) \).

2. The next step involves solving these equations to find an expression for \( \tilde{Q}_{k_1+k_2}^{(1)}(z, \xi) \).

3. Step 3 is to reformulate (B.1) in terms of \( \tilde{Q}_{k_1+k_2}^{(1)}(1, \xi) \). It turns out that in this stage all terms containing probabilities \( \phi_{n,h}^{(1)}(\xi) \) (or their generating functions) are eliminated.

4. The last step involves some more algebraic manipulations which eliminate all terms containing probabilities \( \phi_{n,h}^{(0)}(\xi) \) and, eventually, prove (B.1).

**Step 1: Find relations for the generating functions.** Multiplying Equation (3.16h) with \( z^{n_1} \), summing over all \( n_1 = 1, 2, \ldots \), and adding Equation (3.16b), yields the following equation:

\[
\tilde{Q}_{h_2}^{(1)}(z, \xi) = \tilde{H}(z) \left( \tilde{Q}_{h_2-1}^{(1)}(z, \xi) + \tilde{Q}_{h_2}^{(0)}(z, \xi) - \left( 1 - \frac{\lambda_1}{\mu_1} \right) \tilde{Q}_{h_2}^{(0)}(z, \xi) \right),
\]

(B.2)

for \( h_2 = k_1 + 2, \ldots, k_1 + k_2 \), where \( \tilde{H}(z) \) is defined in (3.8).
Multiplying Equation (3.16g) with $z^{n_1}$, summing over all $n_1 = 1, 2, \ldots$, and adding Equation (3.16a), yields the following equation:

$$
\tilde{Q}^{(1)}_{k_1+1}(z, \xi) = \tilde{H}(z) \left( \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} \phi_{1,h}(\xi) + \frac{\mu_1}{z\mu_2} \tilde{Q}^{(1)}_{k_1}(z, \xi) + \phi_{0,k_1+k_2}(\xi) \right) - \left( 1 - \frac{\lambda_1}{\mu_1} \right) \tilde{Q}^{(0)}_{k_1+1}(z, \xi) .
$$

(B.3)

Multiplying Equation (3.16f) with $z^{n_1+1}$, summing over all $n_1 = 1, 2, \ldots$, and adding Equation (3.16d) multiplied by $z$, yields the following equation:

$$
\tilde{Q}^{(1)}_{h_1}(z, \xi) = \frac{\tilde{G}(z)}{z} \left( \tilde{Q}^{(1)}_{h_1-1}(z, \xi) - \phi_{1,h_1-1}(\xi)z - \frac{z}{\mu_2} \left( 1 - \frac{\lambda_1}{\mu_1} \right) \tilde{Q}^{(0)}_{h_1}(z, \xi) \right) ,
$$

(B.4)

for $h_1 = 2, 3, \ldots, k_1$, where $\tilde{G}(z)$ is defined in (3.8).

Multiplying Equation (3.16e) with $z^{n_1+1}$, summing over all $n_1 = 1, 2, \ldots$, and adding Equation (3.16c) multiplied by $z$, yields the following equation:

$$
\tilde{Q}^{(1)}_{1}(z, \xi) = \frac{\mu_2}{\mu_1} \tilde{G}(z) \left( \tilde{Q}^{(1)}_{k_1+k_2}(z, \xi) - \phi_{0,k_1+k_2}(\xi) + \tilde{Q}^{(0)}_{k_1+k_2}(z, \xi) \right) \left( 1 - \frac{\lambda_1}{\mu_1} \right) \tilde{Q}^{(0)}_{1}(z, \xi) .
$$

(B.5)

**Step 2: Solve these relations and determine $\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi)$**. Solving Equations (B.2)-(B.5) requires a lot of straightforward, but tedious, computations. For reasons of compactness we will present some relevant intermediate results in this appendix, but leave the exact derivations to the reader. Firstly, one can use (B.2), combined with (A.1), to express $\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi)$ in terms of $\tilde{Q}^{(1)}_{k_1+1}(z, \xi)$ and $\tilde{Q}^{(0)}_{k_1+k_2}(z, \xi)$:

$$
\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi) = \tilde{H}(z)^{k_2-1} \tilde{Q}^{(1)}_{k_1+1}(z, \xi) + (k_2 - 1) \left( 1 - \frac{\lambda_1}{\mu_1} \right) \tilde{H}(z) \tilde{Q}^{(0)}_{k_1+k_2}(z, \xi) .
$$

(B.6)

Second, we use (B.3) to express $\tilde{Q}^{(1)}_{k_1+1}(z, \xi)$ in terms of $\tilde{Q}^{(1)}_k(z, \xi)$ and $\tilde{Q}^{(0)}_{k_1+1}(z, \xi)$. Subsequently, we use (B.4) and (A.3) to express $\tilde{Q}^{(1)}_k(z, \xi)$ in terms of $\tilde{Q}^{(1)}_1(z, \xi)$ and $\tilde{Q}^{(0)}_h(z, \xi)$:

$$
\tilde{Q}^{(1)}_k(z, \xi) = \left( \frac{\tilde{G}(z)}{z} \right)^{k_1-1} \tilde{Q}^{(1)}_1(z, \xi) - \sum_{h=1}^{k_1-1} \phi_{1,h}(\xi)z \left( \frac{\tilde{G}(z)}{z} \right)^{k_1-h} \\
- \frac{z}{\mu_2} \left( 1 - \frac{\lambda_1}{\mu_1} \right) \sum_{h=2}^{k_1} \left( \frac{\tilde{G}(z)}{z} \right)^{k_1-h+1} \tilde{Q}^{(0)}_h(z, \xi) .
$$

(B.7)

Finally, we use (B.5) to express $\tilde{Q}^{(1)}_1(z, \xi)$ in terms of $\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi)$ again. After some rearrangement of the terms, this leads to the following expression for $\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi)$:

$$
\tilde{Q}^{(1)}_{k_1+k_2}(z, \xi) = \frac{A^{(1)}(z, \xi) + A^{(0)}(z, \xi)}{D(z)} ,
$$

(B.8)
where
\[
A^{(1)}(z, \xi) = \tilde{H}(z)^{k_2} \left[ \phi_{0,k_1+k_2}^{(1)}(\xi) \left( 1 - \left( \tilde{G}(z)/z \right)^{k_1} \right) + \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} \phi_{1,h}^{(1)}(\xi) \left( 1 - \left( \tilde{G}(z)/z \right)^{k_1-h} \right) \right],
\]
(B.9)
\[
A^{(0)}(z, \xi) = k_2 \tilde{O}_{k_1+k_2}^{(0)}(z, \xi) - \tilde{H}(z)^{k_2} \frac{\mu_1}{\mu_2} \sum_{h=1}^{k_1-1} \phi_{1,h}^{(0)}(\xi) \left( 1 - \left( \tilde{G}(z)/z \right)^{k_1-h} \right)
- \left( 1 - \frac{\lambda_1}{\mu_2} \right) \left[ \tilde{H}(z)^{k_2} \sum_{h=1}^{k_1} \tilde{O}_{h}^{(0)}(z, \xi) \left( \tilde{G}(z)/z \right)^{k_1-h+1} + k_2 \tilde{H}(z) \tilde{O}_{k_1+k_2}^{(0)}(z, \xi) \right],
\]
(B.10)
\[
D(z) = 1 - \left( \tilde{G}(z)/z \right)^{k_1} \tilde{H}(z)^{k_2}.
\]
(B.11)

Note that \(A^{(1)}(z, \xi)\) only contains the probabilities \(\phi_{n,h}^{(1)}(\xi)\). All probabilities \(\phi_{n,h}^{(0)}(\xi)\), and their generating functions, are contained in \(A^{(0)}(z, \xi)\). Also note that \(\frac{A^{(1)}(z, \xi)}{D(z)}\) is exactly the same expression as (3.9), but with constants \(\pi_{n,h}^{(0)}\) replaced by \(\phi_{n,h}^{(0)}\). The reason is that, if one would ignore the probabilities \(\phi_{n,h}^{(0)}(\xi)\) in the balance equations (3.16a)-(3.16h), the system is completely equivalent to the system (3.6a)-(3.6h), which corresponds to the vacation system studied in Appendix A.

Step 3: reformulate the original problem in terms of \(\tilde{O}_{k_1+k_2}^{(1)}(1, \xi)\). In order to solve (B.1), we need to determine \(\sum_{h_1=1}^{k_1} \tilde{O}_{h_1}^{(1)}(1, \xi)\) and \(\sum_{h_1=k_1+1}^{k_1+k_2} \tilde{O}_{h_1}^{(1)}(1, \xi)\). After substituting \(z = 1\) in Equations (B.2)-(B.5), we can express these sums in terms of \(\tilde{O}_{k_1+k_2}^{(1)}(1, \xi)\) and \(\tilde{O}_{h}^{(0)}(1, \xi)\):
\[
\sum_{h_1=1}^{k_1} \tilde{O}_{h_1}^{(1)}(1, \xi) = k_1 \frac{\mu_2}{\mu_1} \left( \tilde{O}_{k_1+k_2}^{(1)}(1, \xi) - \phi_{0,k_1+k_2}^{(1)} \right) - \sum_{h=1}^{k_1-1} (k_1 - h) \phi_{1,h}^{(1)}
+ k_1 \tilde{O}_{1}^{(0)}(1, \xi) - \frac{\mu_2}{\mu_1} \left( 1 - k_1 \right) \frac{\lambda_1}{\mu_1} \sum_{h=1}^{k_1} (k_1 - h + 1) \tilde{O}_{h}^{(0)}(1, \xi),
\]
(B.12)
\[
\sum_{h_1=k_1+1}^{k_1+k_2} \tilde{O}_{h_1}^{(1)}(1, \xi) = k_2 \tilde{O}_{k_1+k_2}^{(1)}(1, \xi) - \frac{\lambda_1}{\mu_1} \left( \frac{k_2-1}{2} \right) \left( 1 - \frac{\lambda_1}{\mu_1} \right) P_{0}^{(1)}(\xi).
\]
(B.13)

Since \(\tilde{O}_{h}^{(0)}(1, \xi)\) for \(h = 1, \ldots, k_1\) can be determined directly using (A.6), it only remains to determine \(\tilde{O}_{k_1+k_2}^{(1)}(1, \xi)\). Fortunately, according to the following lemma we can focus on the part \(\lim_{z \to 1} \frac{A^{(0)}(z, \xi)}{D(z)}\) only.

Lemma B.1 The probabilities \(\phi_{0,k_1+k_2}^{(1)}(1, \xi)\) and \(\phi_{h,k_1}^{(1)}(h = 1, \ldots, k_1 - 1)\) in the left hand side of Equation (B.1) cancel out. Using (B.12) and (B.13), we can express this statement in a more formal presentation:
\[
\lim_{z \to 1} \frac{\lambda_1}{\mu_1} \left( k_2 \frac{A^{(1)}(z, \xi)}{D(z)} \right) - \frac{\mu_2}{\mu_1} \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left( k_1 \frac{\mu_2}{\mu_1} \frac{A^{(1)}(z, \xi)}{D(z)} - \phi_{0,k_1+k_2}^{(1)} \right) - \sum_{h=1}^{k_1-1} (k_1 - h) \phi_{1,h}^{(1)} = 0.
\]
(B.14)
Lemma B.3 Define $X := \lim_{z \to 1} \frac{A^{(0)}(z, \xi)}{D(z)}$. This can be written as

$$X = -\frac{\mu_2}{\mu_1} Y + \left[ k_1 \left( 1 - \frac{\lambda_1}{\mu_1} \right) - k_2 \frac{\lambda_1}{\mu_2} \right]^{-1} \times$$

$$\left\{ \phi^{(0)}_{0,k_1+k_2} \left( k_1 \left( 1 - \frac{\lambda_1}{\mu_1} \right) - k_2 \frac{\lambda_1}{\mu_2} \right) \right. + \left. \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left( k_2 \frac{\lambda_1}{\mu_2} + \frac{\mu_1}{\mu_2} - k_1 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \right) \right\} \phi^{(0)}_{1,h}$$

$$+ \left( 1 - \frac{\lambda_1}{\mu_1} \right) \sum_{h=1}^{k_1-1} h(k_1 - h)\phi^{(0)}_{1,h} - \left( 1 - \frac{\lambda_1}{\mu_1} \right)^2 \left( \frac{k_1 \lambda_1}{\mu_2} - k_1 \left( 1 - \frac{\lambda_1}{\mu_1} \right) \frac{\mu_1}{\mu_2} + \frac{\lambda_1}{\mu_1} \left( \frac{\mu_1}{\mu_2} + k_1 \right) \right) P_0'(\xi). \tag{B.17}$$

Proof:
This equation follows from applying L’Hôpital’s rule to $\frac{A^{(0)}(z, \xi)}{D(z)}$ and, hence, differentiating (B.10) and
Finally, we are ready to present the main result of this appendix, which is the proof of Equation (B.1). Using Lemma B.1, and Equations (B.12) and (B.13), we can write the left hand side of Equation (B.1) as

\[
\frac{\lambda_1}{\mu_1} \left( k_2 X' - \frac{\lambda_1}{\mu_1} \left( k_2 - 1 \right) \left( 1 - \frac{\lambda_1}{\mu_1} \right) P''_0(\xi) \right) \\
- \frac{\lambda_1}{\mu_1} \left( 1 - \frac{\lambda_1}{\mu_1} \right) \left[ k_2 X' + k_1 \widetilde{Q}''_1(1, \xi) - \frac{\mu_2}{\mu_1} \left( 1 - \frac{\lambda_1}{\mu_1} \right) \sum_{h=1}^{k_1} (k_1 - h + 1) \widetilde{Q}''_h(1, \xi) \right],
\]

where \( X' \) is the derivative of \( X \) with respect to \( \xi \). Using Lemma B.3 and Lemma B.2 (and Equation (A.6) to determine \( \widetilde{Q}''_h(1, \xi) \) for \( h = 1, \ldots, k_1 \)) we can show that the above expression reduces to \( \frac{\lambda_1 \mu_1^2}{\mu_1^2} P''_0(\xi) \).

\[\text{References}\]


