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Queue lengths and workloads in polling systems

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ABSTRACT. We consider a polling system: a queueing system of $N \geq 1$ queues with Poisson arrivals $Q_1, \ldots, Q_N$ visited in a cyclic order (with or without switchover times) by a single server. For this system we derive the probability generating function $\mathcal{D}(\cdot)$ of the joint queue length distribution at an arbitrary epoch in a stationary cycle, under no assumptions on service disciplines. We also derive the Laplace-Stieltjes transform $\mathcal{W}(\cdot)$ of the joint workload distribution at an arbitrary epoch. We express $\mathcal{D}$ and $\mathcal{W}$ in the probability generating functions of the joint queue length distribution at visit beginnings, $\mathcal{V}_{bi}(\cdot)$, and visit completions, $\mathcal{V}_{ci}(\cdot)$, at $Q_i$, $i = 1, \ldots, N$. It is well known that $\mathcal{V}_{bi}$ and $\mathcal{V}_{ci}$ can be computed in a broad variety of cases. Furthermore, we establish a workload decomposition result.

1. INTRODUCTION

Consider a queueing model consisting of multiple queues: $Q_1, \ldots, Q_N$, $N \geq 1$, attended by a single server, visiting the queues one at a time in a cyclic order. Moving from one queue to another, the server incurs a (possibly zero) switchover time. Such single-server multiple-queue models are commonly referred to as polling systems. Triggered by a wide variety of applications, polling systems have been extensively studied in the literature, see [11, 14, 16] for a series of comprehensive surveys and [1, 8, 12] for extensive overviews of the applicability of polling systems.

In the polling literature much attention has been given to the determination of the probability generating function (PGF) of the joint queue length distribution of the polling system at various epochs. In particular, the PGF’s of the joint queue length distribution at the epochs at which the server begins a visit, $\mathcal{V}_{bi}(\cdot)$, and completes a visit, $\mathcal{V}_{ci}(\cdot)$, at $Q_i$ were extensively studied. A seminal result has been obtained by Resing [9], who found these PGF’s for polling systems with service disciplines satisfying the so-called branching property. Service disciplines – rules which establish the server’s behaviour while visiting a queue – that satisfy the branching property include the well known exhaustive discipline (per visit the server continues to serve all customers at a station until it empties) and gated discipline (per visit the server serves only those customers at a station which are found there upon its visit).

Polling systems with disciplines which do not satisfy the branching property usually defy exact analysis. A typical example of a discipline that does not satisfy the branching property is the $k$-limited discipline (per visit at most $k$ customers are served at a station). Nevertheless, it is possible to compute $\mathcal{V}_{bi}$ and $\mathcal{V}_{ci}$ for some special cases like completely symmetrical queues or like the two-queue case. For example, for the case $N = 2$ with $Q_1$ served exhaustively and $Q_2$ according to the $k$-limited discipline, the formulas for $\mathcal{V}_{bi}$ and $\mathcal{V}_{ci}$ were found by Winands et al. [17]. Furthermore, for the case of $N = 2$ and $Q_1, Q_2$ both served according to the 1-limited discipline, the formulas for $\mathcal{V}_{bi}$ and $\mathcal{V}_{ci}$ were found by solving a boundary value problem [4]. For other examples of
polling systems with non-branching disciplines that admit a solution for these PGF’s we refer to the surveys [13, 14]. Although the PGF’s \( \mathcal{V}_b \) and \( \mathcal{V}_c \) have been found in a very broad variety of cases, the PGF \( \mathcal{D}(\cdot) \) of the steady-state joint queue length distribution at an arbitrary epoch, which in principle is the most important PGF, received little attention and has not been studied in full generality, one exception (that we are aware of) being Sidi et al. [10]. The first contribution of the present paper is to determine \( \mathcal{D} \) for a quite general class of polling systems. To explain the formula for \( \mathcal{D} \) let us introduce two additional PGF’s. Let \( \mathcal{V}_b(\cdot) \) and \( \mathcal{V}_c(\cdot) \) be the PGF’s of the joint queue length distribution at service beginnings and service completions, respectively. Eisenberg [7] established a link between the various PGF’s: \( \mathcal{V}_b, \mathcal{V}_c, \mathcal{J}_b, \mathcal{J}_c \). In fact, from his observation it follows that it is enough to determine, for example, \( \mathcal{V}_b, i = 1, \ldots, N \), to know all the remaining PGF’s \( \mathcal{V}_c, \mathcal{J}_b, \mathcal{J}_c \), \( i = 1, \ldots, N \). Hence, as noted before, all these PGF’s are well-known for a vast class of polling systems. One of our main results – Theorem 1 – reveals that the formula for \( \mathcal{D} \) we were also able to find \( \mathcal{V}(\cdot) \), the Laplace-Stieltjes transform (LST) of the joint workload distribution in steady state. We give a formula for \( \mathcal{V} \) in Theorem 2 expressed in \( \mathcal{V}_b \) and \( \mathcal{V}_c \), which is the second contribution of our paper. \( \mathcal{D} \) was found in Sidi et al. [10] for polling systems restricted to the case of cycles with either all disciplines being gated or all being exhaustive. \( \mathcal{V} \) was found in [5], for the class of polling systems with Lévy input, but restricted to the case of branching-type service disciplines. Both of the results from [5, 10] assume non-zero switchover times. It should be emphasized therefore that our main results – Theorem 1 and Theorem 2 – hold regardless of the service discipline and provide a unified expression for \( \mathcal{D} \) and \( \mathcal{V} \) for both the zero and non-zero switchover case. These differences are stressed once more in Remark 1 and Remark 4.

From Theorem 2 we were also able to retrieve the well-known decomposition of the total workload \( W \) in a polling system into an independent sum of the amount of work \( W_{M/G/1} \) in a corresponding \( M/G/1 \) queue and \( W_{\text{switch}} \), the total amount of work in the polling system at an arbitrary epoch in a switching period. This decomposition was found in [3], however the distribution of \( W_{\text{switch}} \) was only known in a few cases, cf. [15]. In Remark 5 we present the LST of the distribution of \( W_{\text{switch}} \) expressed in \( \mathcal{V}_b \) and \( \mathcal{V}_c \). This constitutes the third and final contribution of our paper.

The paper is organised as follows. In Section 2 we give a detailed description of the polling model under consideration, introduce the various PGF’s \( \mathcal{V}_b, \mathcal{V}_c, \mathcal{J}_b, \mathcal{J}_c \), and relate them to each other. Section 3 contains our first main result, Theorem 1 – the expression for \( \mathcal{D} \), the PGF of the joint queue length distribution in steady state at an arbitrary epoch, in terms of \( \mathcal{J}_c \). Section 4 contains our second main result, Theorem 2 – the expression for \( \mathcal{V} \), the LST of the joint workload distribution in steady state at an arbitrary epoch, in terms of \( \mathcal{V}_b \) and \( \mathcal{V}_c \). This section also contains Remark 5, which establishes the workload decomposition in polling systems.

2. MODEL DESCRIPTION

2.1. A general polling system. We consider a system of \( N \geq 1 \) infinite-buffer queues \( Q_1, \ldots, Q_N \) and a single server \( S \). The service times of customers in \( Q_i \) are i.i.d. (independent, identically distributed) positive random variables generically denoted by \( B_i \). We denote the LST of \( B_i \) by \( \tilde{B}_i(\cdot) \) and assume that \( b_i := \mathbb{E}B_i < \infty \). The server moves among the queues in a cyclic order. When \( S \) moves from \( Q_i \) to \( Q_{i+1} \), it incurs a switchover period. The durations of successive switchover times are i.i.d. non-negative random variables, which we generically denote by \( S_i \). We denote the LST of \( S_i \) by \( \tilde{S}_i(\cdot) \) and assume that \( s_i := \mathbb{E}S_i < \infty \); let \( s := \sum_{i=1}^{N} s_i \). Customers arrive at \( Q_i \) according to a Poisson process with rate \( \lambda_i \); let \( \lambda := \sum_{i=1}^{N} \lambda_i \). We do not assume anything about the service disciplines at \( Q_i \). Define \( \rho_i := \lambda_i b_i \) as the traffic intensity at \( Q_i \); let \( \rho := \sum_{i=1}^{N} \rho_i \). In what
follows we shall write \( z \) for an \( N \)-dimensional vector in \( \mathbb{R}^N, z = (z_1, \ldots, z_N) \), and we assume that \( |z_i| < 1 \) for every \( i = 1, \ldots, N \). Throughout the paper we implicitly use the convention that any index summation is modulo \( N \), for example \( Q_{N+1} \equiv Q_1 \).

We assume that all the usual independence assumptions hold between the service times, the switchover times and the interarrival times. We assume that the ergodicity conditions are fulfilled and we restrict ourselves to results for the stationary situation.

We also consider the variant of this model in which all the switchover times are zero. In the latter model, \( S \) makes a full cycle (viz., passes all the queues once) when the system becomes empty and subsequently stops right before \( Q_1 \). All this requires zero time. When the first new customer arrives, \( S \) cycles along the queues to that customer. Note that the behavior of the server when the system is empty does not affect the queue length distributions. However, it may involve a number of (zero-length) visits, and hence it does affect the queue length distribution embedded at visit beginning and visit completion instants. The convention that the server stops at some fixed queuing system – which is also a visit beginning epoch. Similarly, all service completion epochs at \( Q_i \) are also service beginning epochs at that queue, except the last service completion epoch – which is also a visit completion epoch.

This has been observed by many authors, but most notably by Eisenberg [7], who restricted himself to the cases of either exhaustive or gated service at all queues. However, its applicability is not restricted to a particular service discipline. After some manipulations this observation yields:

2.2. Distributional identities. For details of this subsection we refer to [2].

Note that each time a visit beginning or a service completion occurs, this coincides with either a service beginning or a visit completion. All service beginning epochs in a visit to \( Q_i \) are also service completion epochs at \( Q_i \), except the first service beginning epoch – which is also a visit beginning epoch. Similarly, all service completion epochs at \( Q_i \) are also service beginning epochs at that queue, except the last service completion epoch – which is also a visit completion epoch. This has been observed by many authors, but most notably by Eisenberg [7], who restricted himself to the cases of either exhaustive or gated service at all queues. However, its applicability is not restricted to a particular service discipline. After some manipulations this observation yields:

\[
\begin{align*}
(1) \quad & \gamma_i \mathcal{V}_b(z) + \mathcal{V}_c(z) = \mathcal{V}_b(z) + \gamma_i \mathcal{V}_c(z). \\
& \text{Here } \mathcal{V}_b(z) \text{ and } \mathcal{V}_c(z) \text{ denote the PGF's of the joint queue length distribution at visit beginnings and visit completions at } Q_i, \text{ while } \mathcal{V}_b(z) \text{ and } \mathcal{V}_c(z) \text{ denote the PGF's of the joint queue length distribution at service beginnings and service completions at } Q_i; \text{ the coefficient } \gamma_i \text{ represents the reciprocal of the mean number of customers served at } Q_i \text{ per visit, i.e., the ratio of visit beginnings to service beginnings. With } C \text{ denoting the steady-state cycle length, } 1/\gamma_i = \lambda_i E_C. \text{ In the case of non-zero switchover times, we simply have } E C = s/(1 - \rho). \text{ In the zero switchover times case we rewrite } (1) \text{ into:}
\end{align*}
\]

\[
\begin{align*}
(2) \quad & \frac{\mathcal{V}_b(0)}{E_C} = \lambda(1 - \rho). \\
& \text{Indeed, } \mathcal{V}_b(0)/EC \text{ denotes the mean number of times per time unit that } S \text{ arrives at } Q_1 \text{ to find the whole system empty (} \mathcal{V}_b(0) \text{ is the probability that } S \text{ sees an empty system at a } Q_1 \text{ visit). Each of those epochs is followed by exactly one arrival of a customer to an empty system. By PASTA, there are on average } \lambda(1 - \rho) \text{ such arrivals per time unit.}
\end{align*}
\]

We rewrite (1) into:

\[
\begin{align*}
(3) \quad & \gamma_i \mathcal{V}_b(z) - \mathcal{V}_c(z) = \mathcal{V}_b(z) - \mathcal{V}_c(z).
\end{align*}
\]

Let \( \Sigma(z) := \sum_{j=1}^N \lambda_j (1 - z_j) \). It is easy to see that \( \mathcal{V}_c(z) \) and \( \mathcal{V}_b(z) \) are related via

\[
\begin{align*}
(4) \quad & \mathcal{V}_c(z) \approx \mathcal{V}_b(z) \frac{\bar{B}_j(\Sigma(z))}{z_j}.
\end{align*}
\]
It follows from (1) and (4) that
\[
\mathcal{X}_b(z) = \frac{\gamma_1 z_i}{z_i - B_i(\Sigma(z))} \left( \mathcal{Y}_{b_i}(z) - \mathcal{Y}_{c_i}(z) \right).
\]

Next we relate \(\mathcal{Y}_{b_{i+1}}\) to \(\mathcal{Y}_{c_i}\). In a polling system with switchover times
\[
\mathcal{Y}_{b_{i+1}}(z) = \mathcal{Y}_{c_i}(z) \tilde{S}_i(\Sigma(z)), \quad i = 1, \ldots, N.
\]

In a polling system without switchover times
\[
\mathcal{Y}_{b_{i+1}}(z) = \mathcal{Y}_{c_i}(z), \quad i = 1, \ldots, N - 1.
\]

The relation between \(\mathcal{Y}_{b_i}\) and \(\mathcal{Y}_{c_N}\) deserves special attention in the zero switchover case, because of our convention stated in Section 2.1, concerning the behavior of the server when the system is empty. When all queues in the model without switchover times become empty, \(S\) in our convention makes a full cycle and subsequently stops right before \(Q_1\) (all this requires no time). When the first new customer arrives, \(S\) cycles along the queues to that customer. The consequence of this is that when the system is empty at the start of a visit to \(Q_1\), then the next visit to \(Q_1\) does not take place until a customer has arrived. Therefore,
\[
\mathcal{Y}_{b_i}(z) = \mathcal{Y}_{c_N}(z) - \mathcal{Y}_{b_i}(0) \left( 1 - \sum_{i=1}^{N} \frac{\lambda_i}{\lambda} z_i \right) = \mathcal{Y}_{c_N}(z) - \frac{\mathcal{Y}_{b_i}(0)}{\lambda} \Sigma(z).
\]

3. Queue Length Distribution at an Arbitrary Time

**Theorem 1.** For a general polling system as introduced in Section 2, let \(\mathcal{D}(\cdot)\) be the probability generating function of the joint queue length distribution at an arbitrary time in steady-state. Then,
\[
\mathcal{D}(z) = \frac{\sum_{i=1}^{N} \lambda_i (1 - z_i) \mathcal{X}_i(z)}{\sum_{i=1}^{N} \lambda_i (1 - z_i)}.
\]

**Proof** We will divide the proof into three parts. First we will find a formula for \(\mathcal{D}\) in the non-zero switchover times case. Next we will show that this formula is also valid in the zero switchover times case. Finally, we will show that the unified formula can be expressed as (9).

Let \(X_i(\cdot)\) and \(Y_i(\cdot)\) be the PGF of the joint queue length distribution at an arbitrary moment during a visit to \(Q_i\) and during a switchover (idle) period between \(Q_i\) and \(Q_{i+1}\), respectively.

**Non-zero switchover times:** By the stochastic mean value theorem
\[
\mathcal{D}(z) = \frac{1}{\mathcal{E}_C} \sum_{i=1}^{N} \left( \frac{b_i}{\gamma_i} X_i(z) + s_i Y_i(z) \right).
\]

Furthermore, for any \(i\),
\[
X_i(z) = \mathcal{X}_b(z) \tilde{B}_i^{\text{past}}(\Sigma(z)) = \mathcal{X}_b(z) \frac{1 - \tilde{B}_i(\Sigma(z))}{\Sigma(z) b_i} = \gamma_i \frac{z_i (\mathcal{Y}_{b_i}(z) - \mathcal{Y}_{c_i}(z)) - \tilde{B}_i(\Sigma(z))}{z_i - \tilde{B}_i(\Sigma(z))}.
\]

where \(\tilde{B}_i^{\text{past}}(\cdot)\) is the LST of \(B_i^{\text{past}}\), the past part of \(B_i\) and where we used (5). Analogously,
\[
Y_i(z) = \mathcal{Y}_{c_i}(z) \tilde{S}_i^{\text{past}}(\Sigma(z)) = \mathcal{Y}_{c_i}(z) \frac{1 - \tilde{S}_i(\Sigma(z))}{\Sigma(z) s_i} = \frac{1}{s_i} \frac{\mathcal{Y}_{c_i}(z) - \mathcal{Y}_{b_{i+1}}(z)}{\Sigma(z)},
\]

where \(\tilde{S}_i^{\text{past}}(\cdot)\) is the LST of \(S_i^{\text{past}}\), the past part of \(S_i\) and where we used (6). Combining (10)-(12) leads to
\[
\mathcal{D}(z) = \frac{1}{\mathcal{E}_C} \sum_{i=1}^{N} \left( \frac{\mathcal{Y}_{b_i}(z) - \mathcal{Y}_{c_i}(z)}{\Sigma(z)} \frac{z_i (1 - \tilde{B}_i(\Sigma(z)))}{z_i - \tilde{B}_i(\Sigma(z))} + \frac{\mathcal{Y}_{c_i}(z) - \mathcal{Y}_{b_{i+1}}(z)}{\Sigma(z)} \right).
\]
Zero switchover times: Observe that in this case we can write
\[ D(z) = \rho D_{\text{serving}}(z) + (1 - \rho) D_{\text{non-serving}}(z), \]
where \( D_{\text{serving}}(\cdot) \) and \( D_{\text{non-serving}}(\cdot) \) denote the PGF of the joint queue length distribution at an arbitrary epoch. However, the result in [10] is restricted to the case of non-zero switchover times.

Remark 1. In Sidi et al. [10] the authors also focus on the joint queue length distribution at an arbitrary epoch. However, the result in [10] is restricted to the case of non-zero switchover times and cycles with either all policies being gated or all being exhaustive. The authors averaged over \( N \) visit times and \( N \) switchover times, but did not obtain our Theorem 1 nor Formula (9).
Our theorem reveals that exactly the same structure holds, regardless of the service discipline. However, they do allow a more general customer behavior; their paper is one of the few polling studies in which the system is viewed as a network, with customers moving from queue to queue and the server visiting the queues cyclically.

Remark 2. Observe that Theorem 1 immediately gives the formula for the marginal distributions. Indeed, for a vector \( z_{MI} = (1, \ldots, 1, z_i, 1, \ldots, 1) \), \( \mathcal{D}(z_{MI}) = \mathcal{D}(z_i) \). From the ‘step’ (level crossing) argument it follows that \( \mathcal{D}(z_i) \) is also the PGF of the queue length distribution in \( Q_i \) at an arrival epoch at \( Q_i \). By PASTA it is also the steady-state distribution of \( Q_i \).

Next take \( z_T = (z, \ldots, z) \). Theorem 1 now states that the PGF of the distribution of the total queue length (in terms of \( z \)) equals \( \sum_{i=1}^{N} \lambda_i \mathcal{D}(z_i) / \sum_{j=1}^{N} \lambda_j \). This formula may be interpreted as follows. By PASTA, \( \mathcal{D}(z_T) \) is also the PGF of the distribution of the total queue length at an arrival epoch. By a level crossing argument, it follows that this equals the PGF of the distribution of the total queue length just after a departure epoch. The result now follows from the observation that a fraction \( \lambda_i / \sum_{j=1}^{N} \lambda_j \) of the departure epochs refers to a departure from \( Q_i \).

Remark 3. It would be very interesting to have an interpretation of (9). Below we present a short derivation, which essentially combines several of the formulas above, but which fails to give an immediate explanation:

For a general polling system as introduced in Section 2, let \( \mathcal{W} \) be the Laplace-Stieltjes transform of the joint workload distribution at an arbitrary time in steady-state. Then,

\[
\mathcal{W}(\omega) = \frac{1}{\mathbb{E}C} \sum_{i=1}^{N} \frac{\mathcal{D}_i(\theta)}{\mathcal{D}(\omega)} \frac{\omega_i}{\Sigma(\theta(\omega)) - \omega_i},
\]

where \( \mathbb{E}C = s/(1 - \rho) \) in the non-zero switchover case and \( \mathbb{E}C = \mathcal{D}_i(0)/(\lambda - \rho) \) in the zero switchover case.

Proof The proof follows the same reasoning as the proof of Theorem 1. In particular, we focus only on the non-zero switchover times case.

Let \( \mathcal{X}_i(\cdot) \) and \( \mathcal{Y}_i(\cdot) \) be the LST of the joint workload distribution at an arbitrary moment during a visit to \( Q_i \) and during a switchover (idle) period between \( Q_i \) and \( Q_{i+1} \), respectively. Then by the
stochastic mean value theorem
\[ \mathcal{W}(\omega) = \frac{1}{BC} \sum_{i=1}^{N} \left( \frac{b_i}{\gamma_i} \dot{X}_i(\omega) + s_i \dot{Y}_i(\omega) \right). \]

Firstly, note that using (12),
\[ \dot{Y}_i(\omega) = Y_i(B(\omega)) = \frac{1}{s_i} \left( \gamma_i Y_i(B(\omega)) - \gamma_i Y_i(B(\omega)) B_i(\omega) - \gamma_i Y_i(B(\omega)) \right). \]

Secondly,
\[ \dot{X}_i(\omega) = \frac{\gamma_i Y_i(B(\omega))}{B_i(\omega)} \times \int_{u=0}^{\infty} \int_{t=0}^{\infty} \exp\left(-\Sigma(B(\omega))u - \omega_i t\right) d\mathbb{P}(B_i^{past} < u, B_i^{res} < t). \]

Indeed, consider an arbitrary service time during a visit to \( Q_i \). The term \( \mathcal{A}_b(B(\omega))/\dot{B}_i(\omega_i) \) corresponds to the LST of the workload of all the customers in the system at the beginning of that service time, excluding the customer whose service is about to begin. Secondly, consider the workload that arrives at all queues during the past part \( B_i^{past} \) of that service time, and the residual part \( B_i^{res} \) of that same service time. Next integrate with respect to the joint distribution of \( B_i^{past} \) and \( B_i^{res} \). Using (5) and the fact that, cf. [6, Section I.6.3],
\[ \int_{u=0}^{\infty} \int_{t=0}^{\infty} \exp\left(-\Sigma(B(\omega))u - \omega_i t\right) d\mathbb{P}(B_i^{past} < u, B_i^{res} < t) = \frac{B_i^{past}(\omega_i) - B_i^{res}(\Sigma(B(\omega)))}{b_i}, \]
we obtain
\[ \dot{X}_i(\omega) = \frac{\gamma_i Y_i(B(\omega))}{B_i(\omega)} \times \frac{B_i^{past}(\omega_i) - B_i^{res}(\Sigma(B(\omega)))}{b_i} = \frac{\gamma_i Y_i(B(\omega))}{B_i(\omega)} \times \frac{B_i^{past}(\omega_i) - B_i^{res}(\Sigma(B(\omega)))}{b_i}. \]

This completes the proof. \( \square \)

Remark 4. \( \mathcal{W} \) was found in [5] for the more general class of polling systems with Lévy input, but restricted to the case of branching-type service disciplines at all queues and non-zero switchover times. Notice that \( \Sigma(B(\omega)) = \omega_i \) is the Laplace exponent \( \phi_A^\omega(\omega) \) of the Lévy process
\[ A_i(t) = (W_1(t), \ldots, W_{i-1}(t), W_i(t) - t, W_{i+1}(t), \ldots, W_N(t)), \]
where \( W_i \) is a compound Poisson process with jump distribution \( B_i \) and rate \( \lambda_i \). Moreover \( \Sigma(B(\omega)) = \omega_i \) is the Laplace exponent of the Lévy process \( W(t) = (W_1(t), \ldots, W_N(t)) \). After identifying \( \gamma_i Y_i(B(\omega)) \) and \( \gamma_i Y_i(B(\omega)) \) of Theorem 2 with \( B_i(\omega) \) and \( \dot{E}_i(\omega) \) of Theorem 3 of [5], it can now be verified that Theorem 2 indeed coincides with [5, Theorem 3].

Remark 5. Observe that Theorem 2 gives the LST of the distribution of \( W \), the total workload in a polling system. With \( \omega_T = (\omega, \ldots, \omega) \), we have \( \mathbb{E}[e^{-\omega T}] = \mathcal{W}(\omega_T) \), so that
\[ \mathbb{E}[e^{-\omega T}] = \frac{1 - \rho}{s} \omega - \sum_{j=1}^{N} \lambda_j (1 - \dot{B}_j(\omega)) \sum_{i=1}^{N} \gamma_i Y_i(B(\omega_T)) - \gamma_i Y_i(B(\omega_T)) \sum_{j=1}^{N} \lambda_j (1 - \dot{B}_j(\omega)). \]

Note that the LST of the amount of work \( W_{M/G/1} \) in a corresponding \( M/G/1 \) queue, that is, an \( M/G/1 \) queue with arrival rate \( \lambda \) and service time LST \( \sum_{j=1}^{N} \lambda_j \dot{B}_j(\omega)/\lambda \), is given by
\[ \mathbb{E}[e^{-\omega T_{M/G/1}}] = \frac{(1 - \rho)\omega}{\omega - \sum_{j=1}^{N} \lambda_j (1 - \dot{B}_j(\omega))}. \]
Note that $W_{M/G/1}$ is also the amount of work in the polling system with zero switchover times. Finally, from the proof of Theorem 2 we know that the LST of the distribution of the total amount of work $W_{\text{switch}}$ in a polling system at an arbitrary epoch in a switching period is given by

$$E[e^{-\omega W_{\text{switch}}}] = \frac{1}{S} \sum_{i=1}^{N} s_i \tilde{Y}_i(\omega_T) = \frac{1}{S} \sum_{i=1}^{N} \frac{\gamma_{ci}(\tilde{B}(\omega_T)) - \gamma_{bi}(\tilde{B}(\omega_T))}{\sum_{j=1}^{N} \lambda_j (1 - B_j(\omega))}.$$  

Hence, we retrieve the well known decomposition (see, e.g., [3])

$$E[e^{-\omega W}] = E[e^{-\omega W_{M/G/1}}] E[e^{-\omega W_{\text{switch}}}] .$$

The distribution of $W_{\text{switch}}$ was only known in a few special cases, cf. [15]. From Theorem 2 it follows that in general its LST is given by (14).

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