# NEGATIVE PROBABILITIES AT WORK IN THE M/D/1 QUEUE 

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This article derives amazingly accurate approximations to the state probabilities and waiting-time probabilities in the $M / D / 1$ queue using a two-phase process with negative probabilities to approximate the deterministic service time. The approximations are in the form of explicit expressions involving geometric and exponential terms. The approximations extend to the finite-capacity $M / D / 1 / N+1$ queue.

## 1. INTRODUCTION

The $M / D / 1$ queue with deterministic services is a simple and generally useful model. It is an appropriate model for telecommunication applications among others (for instance, to describe the cell scale queuing problem in ATM multiplexes). In view of the wide scope of applicability of the $M / D / 1$ model in modern performance evaluation, it is important to have quick and accurate approximations for the state probabilities and the waiting-time probabilities. In this article such approximation will be provided in the form of explicit expressions that are in terms of geometric distributions and exponential densities.

The derivation of the approximation is based on a largely unknown but extremely useful approximation idea of Nojo and Watanabe [4]. It is well known that the probability distribution of a nonnegative random variable with a squared coefficient of variation of at least $\frac{1}{2}$ can be approximated by a Coxian- 2 distribution by matching the first two moments (see, e.g., Tijms [5]). The squared coefficient of variation is the ratio of the variance and the squared mean. In their article, Nojo and Watanabe [4] presented a method to approximate the probability distribution of a nonnegative random variable with a squared coefficient of variation less than $\frac{1}{2}$ by a two-phase distribution with branching probabilities that are negative or larger than one. Using this magical idea, continuous-time Markov chain analysis can be used by doing calculations with negative numbers as if they were legitimate probabilities (see also Van Hoorn and Seelen [6]). The approximation of a random lifetime $X$ through a twophase process is as follows. The process starts in phase 1. It stays in phase 1 for an exponentially distributed time with mean $1 / \gamma$. Upon completion of the sojourn time in phase 1 , the process expires with probability $r_{1}$ and moves to phase 2 with probability $1-r_{1}$. The sojourn time in phase 2 is also exponentially distributed with mean $1 / \gamma$. Upon completion of the sojourn time in phase 2 , the process expires with probability $r_{2}$ and returns to phase 1 with probability $1-r_{2}$. In phase 1 , the process starts anew. The idea is to approximate the original lifetime $X$ by the time it takes until the two-phase process expires. For a random variable $X$ with squared coefficient of variation $c_{X}^{2}<\frac{1}{2}$, it is, under certain side conditions, possible to fit a two-phase process matching the first three moments of $X$ when a negative value or a value larger than one are allowed for the branching probabilities $r_{1}$ and $r_{2}$. These side conditions are satisfied when $X$ has a deterministic distribution. Using the Laplace transform

$$
\begin{equation*}
f^{*}(s)=\frac{\gamma r_{1} s+\gamma^{2}\left(r_{1}+r_{2}-r_{1} r_{2}\right)}{s^{2}+2 \gamma s+\gamma^{2}\left(r_{1}+r_{2}-r_{1} r_{2}\right)} \tag{1}
\end{equation*}
$$

of the density of the expiration time in the two-phase process, it is matter of algebra to derive that

$$
\begin{equation*}
\gamma=\frac{2}{D}, \quad r_{1}=-1, \quad r_{2}=\frac{5}{4} \tag{2}
\end{equation*}
$$

when $X$ is deterministic and equals the constant $D$.
The remainder of the article is organized as follows. In Section 2 we present the explicit expression for the approximations to the state probabilities and the waitingtime probabilities in the $M / D / 1$ queue with infinite waiting room. Approximations for the finite-capacity $M / D / 1 / N+1$ queue are discussed in Section 3. We also give numerical results showing how amazingly accurate the approximations are.

## 2. THE M/D/1 QUEUE

In the single-server $M / D / 1$ queue, the arrival process of customers is a Poisson process with rate $\lambda$ and the service time of any customer is a constant $D$. There is an
infinite waiting room. Customers are served in order of arrival. It is assumed that the offered load,

$$
\rho=\lambda D
$$

is smaller than one. Denoting by $p_{j}(t)$ the probability of having $j$ customers in the system at time $t$, the state probability $p_{j}$ is defined by

$$
p_{j}=\lim _{t \rightarrow \infty} p_{j}(t), \quad j=0,1, \ldots
$$

Also, letting $D_{n}$ denote the delay in queue of the $n$th arrival, the waiting-distribution function $W_{q}(x)$ is defined by

$$
W_{q}(x)=\lim _{n \rightarrow \infty} P\left(D_{n} \leq x\right), \quad x \geq 0
$$

Numerical methods for the exact computation of the $p_{j}$ and $W_{q}(x)$ are available (see, e.g., Tijms [5]). However, for modern performance analysis in telecommunication applications, it is useful to have quick and accurate approximations.

Approximation Result 1: For any $x \geq 0$, the complementary waiting-time distribution function $1-W_{q}(x)$ can be approximated by

$$
\begin{aligned}
& \rho\left[\frac{\rho+2+c(D, \rho)}{2 c(D, \rho)}\right] e^{(1 / 2 D)[\rho-4+c(D, \rho)] x} \\
& \quad-\rho\left[\frac{\rho+2-c(D, \rho)}{2 c(D, \rho)}\right] e^{(1 / 2 D)[\rho-4-c(D, \rho)] x},
\end{aligned}
$$

where $c(D, \rho)$ is the real number $\sqrt{\rho^{2}+16 \rho-8}$ for $\rho \geq-8+6 \sqrt{2}$ and $c(D, \rho)$ is the complex number $i \sqrt{8-16 \rho-\rho^{2}}$ otherwise.

The derivation of this result proceeds as follows. In the general $M / G / 1$ queue, the Laplace transform of the complementary waiting-time distribution $1-W_{q}(x)$ is given by

$$
\begin{equation*}
\int_{0}^{\infty} e^{-s x}\left(1-W_{q}(x)\right) d x=\frac{\rho s-\lambda+\lambda f^{*}(s)}{s\left[s-\lambda+\lambda f^{*}(s)\right]} \tag{3}
\end{equation*}
$$

where $f^{*}(s)$ is the Laplace transform of the service time density (see, e.g., Tijms [5]). Let us now substitute in the right-hand side of (3) the Laplace transform (1) with $\gamma=2 / D, r_{1}=-1$, and $r_{2}=\frac{5}{4}$. In other words, the deterministic service distribution is approximated by the two-phase distribution discussed in Section 1. The right-hand side of (3) then becomes, after some algebra,

$$
\frac{\rho D^{2} s+3 \rho D}{D^{2} s^{2}+(4-\rho) D s+6-6 \rho} .
$$

Denote the numerator of this expression by $N(s)$ and the denominator by $Q(s)$. The Laplace transform $N(s) / Q(s)$ can be analytically inverted, using partial fraction expansion. The (complex) roots of the denominator $Q(s)=D^{2} s^{2}+(4-\rho) D s+$ $6-6 \rho$ are

$$
s_{1,2}=\frac{1}{2 D}\left(\rho-4 \pm \sqrt{\rho^{2}+16 \rho-8}\right) .
$$

Partial fraction expansion gives

$$
\frac{N(s)}{Q(s)}=\frac{N\left(s_{1}\right)}{Q^{\prime}\left(s_{1}\right)} \frac{1}{s-s_{1}}+\frac{N\left(s_{2}\right)}{Q^{\prime}\left(s_{2}\right)} \frac{1}{s-s_{2}} .
$$

Then inversion gives that $N(s) / Q(s)$ is the Laplace transform of

$$
\frac{N\left(s_{1}\right)}{Q^{\prime}\left(s_{1}\right)} e^{s_{1} x}+\frac{N\left(s_{2}\right)}{Q^{\prime}\left(s_{2}\right)} e^{s_{2} x}, \quad x \geq 0
$$

It is then a matter of simple algebra to get Result 1.
For several values of $\rho$, we give in Table 1 the approximate and exact values of $1-W_{q}(x)$ for a number of $x$ values. In all of the examples, the normalization $D=1$ is used. The results in Table 1 show that the approximation is remarkably accurate. The accuracy also applies for small tail probabilities. However, depending on $\rho$, for very small tail probabilities, the approximate value of $1-W_{q}(x)$ might become negative. Fortunately, for very small tail probabilities, one has available the useful asymptotic expansion

$$
1-W_{q}(x) \sim \frac{1-\rho}{\tau \rho-1} e^{-\lambda(\tau-1) x} \quad \text { as } x \rightarrow \infty
$$

where $\tau=1 / \eta$ with $\eta$ the unique root of the equation $\rho(1-x)+x \ln (x)=0$ on $(0,1)$ (see, e.g., Tijms [5, p. 383]). It is interesting to observe that the solution $\eta=\eta(\rho)$ can be very well approximated by the polynomial

$$
\begin{aligned}
\eta_{\text {app }}(\rho)= & -2.4589155 \rho^{9}+12.2135282 \rho^{8}-25.9394588 \rho^{7}+30.8300743 \rho^{6} \\
& -22.5757925 \rho^{5}+10.7001400 \rho^{4}-3.1972952 \rho^{3}+1.2583839 \rho^{2} \\
& +0.1696439 \rho-0.0003216
\end{aligned}
$$

The absolute error of this polynomial approximation is less than $1.1 \times 10^{-5}$ for all practically relevant values of $\rho$ (i.e., for all $\rho \geq 0.2$ ).

Next, we turn to approximations for the state probabilities. For the general $M / G / 1$ queue, the generating function $P(z)=\sum_{j=0}^{\infty} p_{j} z^{j}$ of the state probabilities $p_{j}$ is given by

Table 1. Waiting-Time Probabilities in $M / D / 1$ Queue

| $x$ |  | $\rho=0.2$ | $\rho=0.5$ | $\rho=0.7$ | $\rho=0.8$ | $\rho=0.9$ | $\rho=0.95$ |
| :---: | ---: | ---: | ---: | :--- | :--- | :--- | :--- |
|  | exa | 0.1590 | 0.4334 | 0.6426 | 0.7557 | 0.8748 | 0.9366 |
|  | app | 0.1536 | 0.4244 | 0.6347 | 0.7496 | 0.8712 | 0.9347 |
| 0.5 | exa | 0.1159 | 0.3580 | 0.5743 | 0.7016 | 0.8432 | 0.9196 |
|  | app | 0.1066 | 0.3407 | 0.5579 | 0.6885 | 0.8353 | 0.9153 |
| 1.5 | exa | 0.0085 | 0.1020 | 0.2917 | 0.4553 | 0.6848 | 0.8303 |
|  | app | 0.0109 | 0.1083 | 0.2965 | 0.4579 | 0.6853 | 0.8301 |
| 2.5 | exa | $5.7 \mathrm{E}-4$ | 0.0286 | 0.1477 | 0.2951 | 0.5561 | 0.7496 |
|  | app | $-8.8 \mathrm{E}-4$ | 0.0285 | 0.1500 | 0.2975 | 0.5572 | 0.7500 |
| 3.5 | exa | $3.9 \mathrm{E}-5$ | 0.0081 | 0.0751 | 0.1918 | 0.4520 | 0.6771 |
|  | app | $-3.7 \mathrm{E}-4$ | 0.0070 | 0.0754 | 0.1928 | 0.4529 | 0.6775 |
| 5.5 | exa |  | $6.6 \mathrm{E}-4$ | 0.0195 | 0.0810 | 0.2987 | 0.5525 |
|  | app |  | $3.8 \mathrm{E}-4$ | 0.0190 | 0.0810 | 0.2991 | 0.5527 |
| 7.5 | exa |  | $5.3 \mathrm{E}-5$ | 0.0050 | 0.0342 | 0.1974 | 0.4508 |
|  | app |  | $1.9 \mathrm{E}-5$ | 0.0048 | 0.0340 | 0.1975 | 0.4509 |
| 10.5 | exa |  |  | $6.6 \mathrm{E}-4$ | 0.0094 | 0.1060 | 0.3322 |
|  | app |  |  | $6.0 \mathrm{E}-4$ | 0.0093 | 0.1060 | 0.3323 |
| 15.5 | exa |  |  | $2.3 \mathrm{E}-5$ | 0.0011 | 0.0376 | 0.1998 |
|  | app |  |  | $1.9 \mathrm{E}-5$ | 0.0011 | 0.0376 | 0.1998 |
| 20.5 | exa |  |  |  | $1.3 \mathrm{E}-4$ | 0.0134 | 0.1201 |
|  | app |  |  |  | $1.2 \mathrm{E}-4$ | 0.0133 | 0.1201 |
| 25.5 | exa |  |  |  | $1.5 \mathrm{E}-5$ | 0.0047 | 0.0722 |
|  | app |  |  |  | $1.4 \mathrm{E}-5$ | 0.0047 | 0.0722 |
| 50.5 | exa |  |  |  |  | $2.7 \mathrm{E}-5$ | 0.0057 |
|  | app |  |  |  |  | $2.6 \mathrm{E}-5$ | 0.0057 |

$$
\begin{equation*}
P(z)=\frac{(1-\rho)(1-z) f^{*}(\lambda(1-z))}{f^{*}(\lambda(1-z))-z} \tag{4}
\end{equation*}
$$

where $f^{*}(s)$ is the Laplace transform of the service time density (see, e.g., Tijms [5]). Substituting (1) with $\gamma=2 / D, r_{1}=-1$, and $r_{2}=\frac{5}{4}$ in (4), we get, after some algebra, that

$$
P(z)=\frac{V(z)}{W(z)},
$$

where

$$
\begin{align*}
V(z) & =-2 \rho(1-\rho) z^{2}+(4 \rho-6)(1-\rho) z+(6-2 \rho)(1-\rho)  \tag{5}\\
W(z) & =-\rho^{2} z^{3}+\left(2 \rho^{2}+4 \rho\right) z^{2}-\left(\rho^{2}+2 \rho+6\right) z+6-2 \rho . \tag{6}
\end{align*}
$$

Denoting the complex roots of $W(z)=0$ by $z_{1}, z_{2}$, and $z_{3}$, we then find by partial fraction expansion that

$$
\begin{equation*}
P(z)=\frac{a_{1}}{z-z_{1}}+\frac{a_{2}}{z-z_{2}}+\frac{a_{3}}{z-z_{3}} \tag{7}
\end{equation*}
$$

where

$$
\begin{equation*}
a_{i}=\frac{V\left(z_{i}\right)}{W^{\prime}\left(z_{i}\right)} \quad \text { for } i=1,2,3 \tag{8}
\end{equation*}
$$

The right-hand side of (7) can be inverted as

$$
\frac{a_{1}}{z-z_{1}}+\frac{a_{2}}{z-z_{2}}+\frac{a_{3}}{z-z_{3}}=-\sum_{j=0}^{\infty}\left(\frac{a_{1}}{z_{1}^{j+1}}+\frac{a_{2}}{z_{2}^{j+1}}+\frac{a_{3}}{z_{3}^{j+1}}\right) z^{j} .
$$

Defining the constants $p$ and $q$ by

$$
\begin{aligned}
& p=-\frac{1}{3} \frac{(2 \rho+4)^{2}}{\rho^{2}}+\frac{\rho^{2}+2 \rho+6}{\rho^{2}} \\
& q=-\frac{2}{27} \frac{(2 \rho+4)^{3}}{\rho^{3}}+\frac{1}{3} \frac{2 \rho+4}{\rho} \frac{\rho^{2}+2 \rho+6}{\rho^{2}}+\frac{2 \rho-6}{\rho^{2}},
\end{aligned}
$$

it is a matter of tedious algebra to verify that the roots $z_{1}, z_{2}$, and $z_{3}$ of the equation $W(z)=0$ are given by

$$
\begin{equation*}
z_{i}=w_{i}-\frac{p}{3 w_{i}}+\frac{2 \rho+4}{3 \rho} \quad \text { for } i=1,2,3, \tag{9}
\end{equation*}
$$

where $w_{1}, w_{2}$, and $w_{3}$ are the complex roots of

$$
\begin{equation*}
w^{3}=\frac{1}{2}\left\{-q+\sqrt{q^{2}+\frac{4 p^{3}}{27}}\right\} . \tag{10}
\end{equation*}
$$

Here we used the trick of reducing the equation $x^{3}+p x+q=0$ to the equation $w^{6}+q w^{3}-\left(p^{3} / 27\right)=0$ by the change of variable $x=w-(p / 3 w)$. It is standard to solve (10). The complex roots of the equation $u^{3}=1$ are calculated as

$$
\begin{equation*}
u_{1}=1, \quad u_{2}=e^{i 2 \pi / 3}, \quad u_{3}=e^{-i 2 \pi / 3} \tag{11}
\end{equation*}
$$

Summarizing, we have the following result.
Approximation Result 2: For any $j=0,1, \ldots$, the state probability $p_{j}$ in the $M / D / 1$ queue can be approximated by

$$
-a_{1} z_{1}^{-(j+1)}-a_{2} z_{2}^{-(j+1)}-a_{3} z_{3}^{-(j+1)}
$$

where $z_{1}, z_{2}$, and $z_{3}$ are calculated from (9)-(11) and the constants $a_{1}, a_{2}$, and $a_{3}$ are calculated from (8).

The calculations for the approximate $p_{j}$ are simple and offer no numerical difficulties, unlike the calculations with the algebraic expression for the exact $p_{j}$ in Brun and Garcia [1]. For several values of $\rho$, we give in Table 2 the approximate and exact values of the tail probabilities

$$
Q_{j}=\sum_{k=j+1}^{\infty} p_{k} .
$$

The results in Table 2 show that the approximations are remarkably accurate. The accuracy also holds for small tail probabilities. However, for very small values of $Q_{j}$, it is better to use the asymptotic expansion

$$
Q_{j} \sim \frac{1-\rho}{\tau \rho-1} \tau^{-j} \quad \text { as } j \rightarrow \infty
$$

Table 2. The State Probabilities $Q_{j}$ in the $M / D / 1$ Queue

| $j$ |  | $\rho=0.2$ | $\rho=0.5$ | $\rho=0.7$ | $\rho=0.8$ | $\rho=0.9$ | $\rho=0.95$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 0 | exa | 0.2000 | 0.5000 | 0.7000 | 0.8000 | 0.9000 | 0.9500 |
|  | app | 0.2000 | 0.5000 | 0.7000 | 0.8000 | 0.9000 | 0.9500 |
| 1 | exa | 0.0229 | 0.1756 | 0.3959 | 0.5549 | 0.7540 | 0.8707 |
|  | app | 0.0229 | 0.1750 | 0.3941 | 0.5527 | 0.7521 | 0.8695 |
| 2 | exa | 0.0020 | 0.0530 | 0.2063 | 0.3655 | 0.6164 | 0.7885 |
|  | app | 0.0020 | 0.0538 | 0.2073 | 0.3660 | 0.6162 | 0.7881 |
| 3 | exa | $1.5 \mathrm{E}-4$ | 0.0153 | 0.1053 | 0.2379 | 0.5013 | 0.7124 |
|  | app | $1.4 \mathrm{E}-4$ | 0.0154 | 0.1063 | 0.2390 | 0.5019 | 0.7126 |
| 5 | exa |  | 0.0012 | 0.0273 | 0.1005 | 0.3313 | 0.5813 |
|  | app |  | 0.0011 | 0.0273 | 0.1008 | 0.3318 | 0.5815 |
| 7 | exa |  | $1.0 \mathrm{E}-4$ | 0.0071 | 0.0425 | 0.2189 | 0.4743 |
|  | app |  | $7.9 \mathrm{E}-5$ | 0.0069 | 0.0424 | 0.2192 | 0.4745 |
| 10 | exa |  |  | $9.3 \mathrm{E}-4$ | 0.0117 | 0.1176 | 0.3495 |
|  | app |  |  | $8.9 \mathrm{E}-4$ | 0.0116 | 0.1176 | 0.3497 |
| 15 | exa |  |  | $3.2 \mathrm{E}-5$ | 0.0014 | 0.0417 | 0.2102 |
|  | app |  |  | $2.9 \mathrm{E}-5$ | 0.0013 | 0.0417 | 0.2102 |
| 20 | exa |  |  |  | $1.6 \mathrm{E}-4$ | 0.0148 | 0.1264 |
|  | app |  |  |  | $1.5 \mathrm{E}-4$ | 0.0148 | 0.1264 |
| 25 | exa |  |  |  | $1.8 \mathrm{E}-5$ | 0.0053 | 0.0760 |
|  | app |  |  |  | $1.7 \mathrm{E}-5$ | 0.0052 | 0.0760 |
| 50 | exa |  |  |  |  | $3.0 \mathrm{E}-5$ | 0.0060 |
|  | app |  |  |  |  | $2.9 \mathrm{E}-5$ | 0.0060 |

where $\tau$ is the same as in the asymptotic expansion of $1-W_{q}(x)$ (see Tijms [5, p. 378]).

## 3. THE $M / D / 1 / N+1$ QUEUE

The $M / D / 1 / N+1$ queue differs only from the $M / D / 1$ queue by having a finite waiting room with capacity $N$ rather than an infinite waiting room. An arriving customer who finds, upon arrival, the server busy and all of the $N$ additional waiting places occupied is rejected and has no further influence on the system. In the discussion below, it is again assumed that $\rho=\lambda D$ is smaller than one. Under this assumption, the stationary state probabilities $p_{j}^{(N)}$ in the $M / D / 1 / N+1$ queue can be expressed in the corresponding state probabilities $p_{j}$ in the $M / D / 1$ queue. The state probabilities $p_{j}^{(N)}$ satisfy

$$
p_{j}^{(N)}= \begin{cases}\gamma p_{j} & \text { for } j=0,1, \ldots, N, \\ 1-\sum_{k=0}^{N} p_{k}^{(N)} & \text { for } j=N+1\end{cases}
$$

where the proportionality constant $\gamma$ is given by

$$
\gamma=\left[1-\rho\left(1-\sum_{j=0}^{N} p_{j}\right)\right]^{-1}
$$

(see, e.g., Tijms [5, p. 410]). Thus, Approximation Result 2 leads directly to simple approximations for the state probabilities $p_{j}^{(N)}$. In turn, the state probabilities $p_{j}^{(N)}$ can be used to compute the stationary waiting-time distribution function $W_{q}(x)$, using a beautiful result of Franx [2,3]. In Franx [2], a simple and numerically stable method was given to calculate the waiting-time probabilities in the multiserver $M / D / c$ queue with infinite capacity. This exact method can be modified for the finite-capacity model when $c=1$. For the $M / D / 1 / N+1$ model, let us define $W_{q}(x)=\lim _{n \rightarrow \infty} P\left(D_{n} \leq x\right)$, with $D_{n}$ denoting the delay in queue of the $n$th accepted customer. The following exact result holds for the $M / D / 1 / N+1$ queue. For $(k-1) D \leq x<k D$ and $k=1, \ldots, N$, the waiting-time probability $W_{q}(x)$ is given by

$$
W_{q}(x)=\frac{1}{1-p_{N+1}^{(N)}} \sum_{j=0}^{k-1} Q_{k-1-j} e^{-\lambda(k D-x)} \frac{[\lambda(k D-x)]^{j}}{j!},
$$

where $Q_{r}=\sum_{i=0}^{r+1} p_{i}^{(N)}$ for $r=0,1, \ldots, N$. This exact formula applies to the case of $\rho \geq 1$ as well. For the case of $\rho<1$, we can use the simple approximations for the $p_{j}^{(N)}$ in order to get an easily calculated approximation for $W_{q}(x)$.

In Table 3 we show that Approximation Result 2 leads to excellent approximations to the $p_{j}^{(N)}$. For several values of $\rho$ and $N$, we give the approximate and

Table 3. Various Performance Measures for the $M / D / 1 / N+1$ Queue

| $\rho$ |  | $N=2$ |  |  |  | $N=5$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $P_{\text {rej }}$ | $P_{\text {delay }}$ | $E\left(L_{q}\right)$ | $\sigma\left(L_{q}\right)$ | $P_{\text {rej }}$ | $P_{\text {delay }}$ | $E\left(L_{q}\right)$ | $\sigma\left(L_{q}\right)$ |
| 0.5 | exa | 0.0272 | 0.4720 | 0.1804 | 0.4498 | 0.0006 | 0.4994 | 0.2466 | 0.6136 |
|  | app | 0.0276 | 0.4716 | 0.1798 | 0.4503 | 0.0006 | 0.4994 | 0.2469 | 0.6153 |
| 0.8 | exa | 0.1033 | 0.6848 | 0.4743 | 0.6752 | 0.0219 | 0.7777 | 1.0998 | 1.3618 |
|  | app | 0.1035 | 0.6845 | 0.4710 | 0.6754 | 0.0219 | 0.7776 | 1.0993 | 1.3652 |
| 0.9 | exa | 0.1384 | 0.7393 | 0.5860 | 0.7208 | 0.0472 | 0.8505 | 1.5955 | 1.5580 |
|  | app | 0.1383 | 0.7395 | 0.5819 | 0.7211 | 0.0473 | 0.8503 | 1.5928 | 1.5610 |
| 0.95 | exa | 0.1571 | 0.7636 | 0.6419 | 0.7377 | 0.0649 | 0.8806 | 1.8656 | 1.6169 |
|  | app | 0.1568 | 0.7640 | 0.6374 | 0.7381 | 0.0650 | 0.8805 | 1.8617 | 1.6199 |
| 0.99 | exa | 0.1725 | 0.7816 | 0.6860 | 0.7486 | 0.0813 | 0.9015 | 2.0839 | 1.6425 |
|  | app | 0.1720 | 0.7823 | 0.6814 | 0.7491 | 0.0813 | 0.9015 | 2.0793 | 1.6455 |

exact values of the rejection probability $P_{\text {rej }}=p_{N+1}^{(N)}$, the delay probability $P_{\text {delay }}=$ $\sum_{j=1}^{N} p_{j}^{(N)} /\left(1-p_{N+1}^{(N)}\right), E\left(L_{q}\right)$, and $\sigma\left(L_{q}\right)$, where $E\left(L_{q}\right)$ and $\sigma\left(L_{q}\right)$ denote the expected value and the standard deviation of the stationary queue size $L_{q}$.

For several values of $\rho$ and $N$, we give in Table 4 the approximate and exact values of $1-W_{q}(x)$ for a number of $x$ values. The results in Table 4 show an excellent performance of the approximation.

Summarizing, we can conclude for the $M / D / 1 / N+1$ queue that approximating the deterministic service time distribution by a two-phase process with a negative

Table 4. Results for $1-W_{q}(x)$ in the $M / D / 1 / N+1$ Queue

| $\rho$ |  | $N=2$ |  |  |  | $N=5$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $x=0.5$ | 0.75 | 1.5 | 1.9 | $x=0.5$ | 1.75 | 3.9 | 4.5 |
| 0.5 | exa | 0.3220 | 0.2318 | 0.0517 | 0.0074 | 0.3572 | 0.0722 | 0.0037 | 0.0011 |
|  | app | 0.3210 | 0.2306 | 0.0514 | 0.0073 | 0.3568 | 0.0729 | 0.0037 | 0.0011 |
| 0.8 | exa | 0.5298 | 0.4257 | 0.1416 | 0.0251 | 0.6683 | 0.3414 | 0.0677 | 0.0269 |
|  | app | 0.5271 | 0.4224 | 0.1405 | 0.0248 | 0.6666 | 0.3413 | 0.0683 | 0.0270 |
| 0.9 | exa | 0.5912 | 0.4880 | 0.1784 | 0.0333 | 0.7655 | 0.4759 | 0.1268 | 0.0541 |
|  | app | 0.5882 | 0.4843 | 0.1771 | 0.0330 | 0.7635 | 0.4747 | 0.1271 | 0.0542 |
| 0.95 | exa | 0.6198 | 0.5179 | 0.1975 | 0.0378 | 0.8080 | 0.5439 | 0.1644 | 0.0724 |
|  | app | 0.6169 | 0.5142 | 0.1961 | 0.0374 | 0.8060 | 0.5423 | 0.1644 | 0.0724 |
| 0.99 | exa | 0.6417 | 0.5411 | 0.2131 | 0.0415 | 0.8384 | 0.5966 | 0.1976 | 0.0893 |
|  | app | 0.6388 | 0.5374 | 0.2116 | 0.0411 | 0.8365 | 0.5948 | 0.1975 | 0.0892 |

branching probability leads to remarkably accurate approximations that are easy to use in practical applications. The approximation approach works only for the steadystate behavior of the system, not for the transient behavior. It is our conjecture that the approximation approach in conjunction with numerical Laplace inversion will lead to practically useful approximations for the stationary waiting-time probabilities in the notoriously difficult multiserver $M / D / c / N+c$ queue. This will be the subject of further research.

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