

Moments of Integrated Work in the M/G/1

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This note gives a straightforward way to compute the moments of the distribution of the integrated amount of work over the course of a busy period in an M/G/1. This quantity is the area under the work function, over a busy period. This problem was posed by Iglehart [3], with further work by Cohen [1] and Glynn et al. [2].

Consider an M/G/1 with arrival rate λ and job size distribution with random variable S . Let the state of the M/G/1 be (a, w) , where a is the integrated work in the busy period so far, and w is the work in the system. During an idle period, a will remain at its value at the end of the previous busy period, resetting to 0 at the beginning of the next busy period.

Let A and W be the time-average values of a and w , and let A_r be the value of A at reset points.

The goal of this note is to determine the distribution of A_r . Specifically, we will calculate all of the moments of A_r .

1 Drift Method

There are four events in the system that can affect the drift: The deterministic increase in a , the deterministic decrease in w , arrivals, and resets at the beginning of busy periods.

w decreases at rate 1, with stochastic jumps of size S at rate λ . a increases at rate w , resetting to 0 at the beginning of each busy period.

We can thus write down the drift for an arbitrary test function $f(a, w)$. Let G denote the instantaneous generator of the M/G/1 system:

$$G \circ f(a, w) = w \frac{\partial f(a, w)}{\partial a} - \frac{\partial f(a, w)}{\partial w} + \lambda(E[f(a, w + S)] - f(a, w))\mathbb{1}\{w > 0\} - \lambda(f(a, 0) - E[f(0, S)])\mathbb{1}\{w = 0\}$$

Then, we can take expectation and apply the fact that $E[G \circ f(A, W)] = 0$.

To determine $E[A_r^k]$ for all k , we will use the drift method with test functions $a^k w^\ell$, $k, \ell \geq 0$. There are three cases to split things into: $k = 0, \ell = 0, k, \ell > 0$.

1.1 Area only ($\ell = 0$)

First, we compute the drift:

$$G \circ a^k = ka^{k-1}w - \lambda a^k \mathbb{1}\{w = 0\}$$

Next, we take expectations:

$$\begin{aligned} 0 &= kE[A^{k-1}W] - \lambda(1 - \rho)E[A_r^k] \\ E[A_r^k] &= \frac{kE[A^{k-1}W]}{\lambda(1 - \rho)} \end{aligned}$$

Note that we make use of the fact that arrivals are Poisson, and the PASTA principle, in this step. This ensures that $E[A | W = 0]$ and $E[A_r]$ are equal.

Thus, to compute moments of A_r , it suffices to compute expectations of multivariate polynomials of A and W .

1.2 Area and Work ($k, \ell > 0$)

First, we compute the drift:

$$\begin{aligned} G \circ a^k w^\ell &= ka^{k-1}w^{\ell+1} - \ell a^k w^{\ell-1} + \lambda a^k (E[(w + S)^\ell] - w^\ell) \\ &= ka^{k-1}w^{\ell+1} - \ell a^k w^{\ell-1} + \lambda a^k \sum_{i=1}^{\ell} \binom{\ell}{i} E[S^i] w^{\ell-i} \\ &= -(1 - \rho)\ell(a^k w^{\ell-1}) + ka^{k-1}w^{\ell+1} + \lambda a^k \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] w^{\ell-i} \end{aligned}$$

Next, we take expectations:

$$\begin{aligned} 0 &= -(1 - \rho)\ell E[A^k W^{\ell-1}] + kE[A^{k-1}W^{\ell+1}] + \lambda \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] E[A^k W^{\ell-i}] \\ E[A^k W^{\ell-1}] &= \frac{kE[A^{k-1}W^{\ell+1}] + \lambda \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] E[A^k W^{\ell-i}]}{\ell(1 - \rho)} \end{aligned}$$

Note that the expectation of a multivariate polynomial of A and W can be expressed in terms of expectations of multivariate polynomials with either the same total degree and lower degree of A , or lower total degree and the same degree of A .

Thus, to compute these expectations, it suffices to compute moments of W .

1.3 Work only ($k = 0$)

First, we compute the drift:

$$\begin{aligned} G \circ w^\ell &= -\ell w^{\ell-1} + \lambda \sum_{i=1}^{\ell} \binom{\ell}{i} E[S^i] w^{\ell-i} \\ &= -\ell(1-\rho)w^{\ell-1} + \lambda \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] w^{\ell-i} \end{aligned}$$

Next, we take expectations:

$$\begin{aligned} 0 &= -\ell(1-\rho)E[W^{\ell-1}] + \lambda \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] E[W^{\ell-i}] \\ E[W^{\ell-1}] &= \frac{\lambda \sum_{i=2}^{\ell} \binom{\ell}{i} E[S^i] E[W^{\ell-i}]}{\ell(1-\rho)} \end{aligned}$$

Note that the moments of W are expressed in terms of lower-degree moments of W and the moments of S . Thus, we can express the moments of A_r in terms of the moments of the size S .

2 Examples

To demonstrate the method, I will compute the first, second, and third moments of A_r , given in (1), (2), (3).

2.1 First moment

$$\begin{aligned} E[A_r] &= \frac{E[W]}{\lambda(1-\rho)} \\ E[W] &= \frac{\lambda E[S^2]}{2(1-\rho)} \\ E[A_r] &= \frac{E[S^2]}{2(1-\rho)^2} \end{aligned} \tag{1}$$

Note that (1) matches Glynn et al. [2]'s equation (20), for which they cite Iglehart [3]'s original paper, for which I believe the relevant result is Lemma 2.4(c), though I'm not certain – the paper uses very different notation. This also matches Cohen [1]'s equation (3.4).

2.2 Second moment

$$\begin{aligned}
E[A_r^2] &= \frac{2E[AW]}{\lambda(1-\rho)} \\
E[AW] &= \frac{E[W^3] + \lambda E[S^2]E[A]}{2(1-\rho)} \\
E[A] &= \frac{E[W^2]}{1-\rho} \\
E[A_r^2] &= \frac{E[W^3]}{\lambda(1-\rho)^2} + \frac{E[S^2]E[A]}{(1-\rho)^2} = \frac{E[W^3]}{\lambda(1-\rho)^2} + \frac{E[W^2]E[S^2]}{(1-\rho)^3} \\
E[W^2] &= \frac{3\lambda E[S^2]E[W] + \lambda E[S^3]}{3(1-\rho)} = \frac{\lambda^2 E[S^2]^2}{2(1-\rho)^2} + \frac{\lambda E[S^3]}{3(1-\rho)} \\
E[W^3] &= \frac{6\lambda E[S^2]E[W^2] + 4\lambda E[S^3]E[W] + \lambda E[S^4]}{4(1-\rho)} \\
&= \frac{3\lambda^3 E[S^2]^3}{4(1-\rho)^3} + \frac{\lambda^2 E[S^2]E[S^3]}{(1-\rho)^2} + \frac{\lambda E[S^4]}{4(1-\rho)} \\
E[A_r^2] &= \frac{5\lambda^2 E[S^2]^3}{4(1-\rho)^5} + \frac{4\lambda E[S^2]E[S^3]}{3(1-\rho)^4} + \frac{E[S^4]}{4(1-\rho)^3} \tag{2}
\end{aligned}$$

Note that (2) matches Cohen [1]'s equation (3.9). Note that it does not match Glynn et al. [2]'s equation (21), despite that paper citing Cohen [1] as its source for the equation. I believe this is a transcription error – the 4 in the denominator of the $E[S^2]^3$ term and the 3 in the denominator of the $E[S^2]E[S^3]$ term are missing in Glynn et al. [2].

2.3 Third moment

$$\begin{aligned}
E[A_r^3] &= \frac{3E[A^2W]}{\lambda(1-\rho)} \\
E[A^2W] &= \frac{2E[AW^3] + \lambda E[S^2]E[A^2]}{2(1-\rho)} \\
E[A^2] &= \frac{2E[AW^2]}{1-\rho} \\
E[AW^3] &= \frac{E[W^5] + 6\lambda E[S^2]E[AW^2] + 4\lambda E[S^3]E[AW] + \lambda E[S^4]E[A]}{4(1-\rho)} \\
E[AW^2] &= \frac{E[W^4] + 3\lambda E[S^2]E[AW] + \lambda E[S^3]}{3(1-\rho)}
\end{aligned}$$

$$\begin{aligned}
E[A_r^3] &= \frac{3E[AW^3]}{\lambda(1-\rho)^2} + \frac{3E[S^2]E[A^2]}{2(1-\rho)^2} \\
&= \frac{3E[W^5]}{4\lambda(1-\rho)^3} + \frac{15E[S^2]E[AW^2]}{2(1-\rho)^3} + \frac{3E[S^3]E[AW]}{(1-\rho)^3} + \frac{3E[S^4]E[A]}{4(1-\rho)^3} \\
&= \frac{3E[W^5]}{4\lambda(1-\rho)^3} + \frac{5E[W^4]E[S^2]}{2(1-\rho)^4} + E[AW] \left(\frac{15\lambda E[S^2]^2}{2(1-\rho)^4} + \frac{3E[S^3]}{(1-\rho)^3} \right) \\
&\quad + \frac{3E[S^4]E[A]}{4(1-\rho)^3} \\
&= \frac{3E[W^5]}{4\lambda(1-\rho)^3} + \frac{5E[W^4]E[S^2]}{2(1-\rho)^4} + E[W^3] \left(\frac{15\lambda E[S^2]^2}{4(1-\rho)^5} + \frac{3E[S^3]}{2(1-\rho)^4} \right) \\
&\quad + E[A] \left(\frac{15\lambda^2 E[S^2]^3}{4(1-\rho)^5} + \frac{3\lambda E[S^3]E[S^2]}{2(1-\rho)^4} + \frac{3E[S^4]}{4(1-\rho)^3} \right) \\
E[A_r^3] &= \frac{3E[W^5]}{4\lambda(1-\rho)^3} + \frac{5E[W^4]E[S^2]}{2(1-\rho)^4} + E[W^3] \left(\frac{15\lambda E[S^2]^2}{4(1-\rho)^5} + \frac{3E[S^3]}{2(1-\rho)^4} \right) \\
&\quad + E[W^2] \left(\frac{15\lambda^2 E[S^2]^3}{4(1-\rho)^6} + \frac{3\lambda E[S^3]E[S^2]}{2(1-\rho)^5} + \frac{3E[S^4]}{4(1-\rho)^4} \right) \\
E[W^5] &= \frac{15\lambda E[S^2]E[W^4] + 20\lambda E[S^3]E[W^3] + 15\lambda E[S^4]E[W^2] + 6\lambda E[S^5]E[W] + \lambda E[S^6]}{6(1-\rho)} \\
E[W^4] &= \frac{10\lambda E[S^2]E[W^3] + 10\lambda E[S^3]E[W^2] + 5\lambda E[S^4]E[W] + \lambda E[S^5]}{5(1-\rho)}
\end{aligned}$$

$$\begin{aligned}
E[A_r^3] &= \frac{35E[W^4]E[S^2]}{8(1-\rho)^4} + E[W^3] \left(\frac{15\lambda E[S^2]^2}{4(1-\rho)^5} + \frac{4E[S^3]}{(1-\rho)^4} \right) \\
&\quad + E[W^2] \left(\frac{15\lambda^2 E[S^2]^3}{4(1-\rho)^6} + \frac{3\lambda E[S^3]E[S^2]}{2(1-\rho)^5} + \frac{21E[S^4]}{8(1-\rho)^4} \right) + \frac{3E[W]E[S^5]}{4(1-\rho)^4} + \frac{E[S^6]}{8(1-\rho)^4} \\
&= E[W^3] \left(\frac{25\lambda E[S^2]^2}{2(1-\rho)^5} + \frac{4E[S^3]}{(1-\rho)^4} \right) \\
&\quad + E[W^2] \left(\frac{15\lambda^2 E[S^2]^3}{4(1-\rho)^6} + \frac{41\lambda E[S^3]E[S^2]}{4(1-\rho)^5} + \frac{21E[S^4]}{8(1-\rho)^4} \right) \\
&\quad + E[W] \left(\frac{35\lambda E[S^4]E[S^2]}{8(1-\rho)^5} + \frac{3E[S^5]}{4(1-\rho)^4} \right) + \frac{35\lambda E[S^5]E[S^2]}{8(1-\rho)^5} + \frac{E[S^6]}{8(1-\rho)^4} \\
&= E[W^2] \left(\frac{45\lambda^2 E[S^2]^3}{2(1-\rho)^6} + \frac{65\lambda E[S^3]E[S^2]}{4(1-\rho)^5} + \frac{21E[S^4]}{8(1-\rho)^4} \right) \\
&\quad + E[W] \left(\frac{25\lambda^2 E[S^3]E[S^2]^2}{2(1-\rho)^6} + \frac{4\lambda E[S^3]^2}{(1-\rho)^5} + \frac{35\lambda E[S^4]E[S^2]}{8(1-\rho)^5} + \frac{3E[S^5]}{4(1-\rho)^4} \right) \\
&\quad + \frac{25\lambda^2 E[S^4]E[S^2]^2}{8(1-\rho)^6} + \frac{\lambda E[S^4]E[S^3]}{(1-\rho)^5} + \frac{35\lambda E[S^5]E[S^2]}{8(1-\rho)^5} + \frac{E[S^6]}{8(1-\rho)^4}
\end{aligned}$$

$$\begin{aligned}
E[A_r^3] &= E[W] \left(\frac{45\lambda^3 E[S^2]^4}{2(1-\rho)^7} + \frac{115\lambda^2 E[S^3]E[S^2]^2}{4(1-\rho)^6} + \frac{4\lambda E[S^3]^2}{(1-\rho)^5} + \frac{7\lambda E[S^4]E[S^2]}{(1-\rho)^5} + \frac{3E[S^5]}{4(1-\rho)^4} \right) \\
&+ \frac{15\lambda^3 E[S^3]E[S^2]^3}{2(1-\rho)^7} + \frac{65\lambda^2 E[S^3]^2 E[S^2]}{12(1-\rho)^6} + \frac{25\lambda^2 E[S^4]E[S^2]^2}{8(1-\rho)^6} \\
&+ \frac{15\lambda E[S^4]E[S^3]}{8(1-\rho)^5} + \frac{35\lambda E[S^5]E[S^2]}{8(1-\rho)^5} + \frac{E[S^6]}{8(1-\rho)^4} \\
E[A_r^3] &= \frac{45\lambda^4 E[S^2]^5}{4(1-\rho)^8} + \frac{175\lambda^3 E[S^3]E[S^2]^3}{8(1-\rho)^7} + \frac{89\lambda^2 E[S^3]^2 E[S^2]}{12(1-\rho)^6} + \frac{53\lambda^2 E[S^4]E[S^2]^2}{8(1-\rho)^6} \\
&+ \frac{15\lambda E[S^4]E[S^3]}{8(1-\rho)^5} + \frac{19\lambda E[S^5]E[S^2]}{4(1-\rho)^5} + \frac{E[S^6]}{8(1-\rho)^4} \tag{3}
\end{aligned}$$

To the best of my knowledge (and that of Glynn et al. [2]), (3) has not previously appeared in the literature.

I will stop here, but such formulas for arbitrary moments can be computed by this method with the aid of a computer algebra system.

References

- [1] JW Cohen. Properties of the process of level crossings during a busy cycle of the M/G/1 queueing system. *Mathematics of Operations Research*, 3(2): 133–144, 1978.
- [2] Peter W Glynn, Royi Jacobovic, and Michel Mandjes. Moments of polynomial functionals in Lévy-driven queues with secondary jumps. *arXiv preprint arXiv:2310.11137*, 2023.
- [3] Donald L Iglehart. Functional limit theorems for the queue GI/G/1 in light traffic. *Advances in Applied Probability*, 3(2):269–281, 1971.