



ON TRANSCIENCE OF M/G/∞ QUEUES

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Abstract

We consider an M/G/∞ queue with infinite expected service time. We then provide the transience/recurrence classification of the states (the system is said to be at state n if there are n customers being served), observing also that here (unlike irreducible Markov chains, for example) it is possible for recurrent and transient states to coexist. We also prove a lower bound on the growth speed in the transient case.

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In this note we consider a classical M/G/∞ queue (see e.g. [3]): the customers arrive according to a Poisson process with rate λ ; upon arrival, a customer immediately enters the service, and the service times are i.i.d. (non-negative) random variables with some general distribution. For notational convenience, let S be a generic random variable with that distribution. We also assume that at time 0 there are no customers being served. Let Y_t denote the number of customers in the system at time t , which we also refer to as the *state of the system* at time t ; note that, in general, Y is not a Markov process. Nevertheless, let us still call a state m *recurrent* if the set $\{t : Y_t = m\}$ is a.s. unbounded (i.e. m is ‘visited infinitely many times’) and *transient* if that set is a.s. bounded (i.e. the system escapes from m eventually). Let us also observe that $\mathbb{P}[S = \infty] > 0$ clearly implies that Y goes to infinity (indeed, it is straightforward to obtain that $\liminf_{t \rightarrow \infty} (Y_t/t) \geq \lambda \mathbb{P}[S = \infty]$), so from now on we assume that $S < \infty$ a.s.

We are mainly interested in the situation where the system is *unstable*, i.e. when $\mathbb{E}S = \infty$. In this situation, in principle, our (Markovian) intuition tells us that the system can be *transient* (in the sense $Y_t \rightarrow \infty$ a.s.) or recurrent (i.e. all states are visited infinitely many times a.s.). However, it turns out that for this model the complete picture is more complicated.

Theorem 1. *Define*

$$k_0 = \min \left\{ k \in \mathbb{Z}_+ : \int_0^\infty (\mathbb{E}(S \wedge t))^k \exp(-\lambda \mathbb{E}(S \wedge t)) dt = \infty \right\} \quad (1)$$

(with the convention $\min \emptyset = +\infty$). Then

$$\liminf_{t \rightarrow \infty} Y_t = k_0 \quad a.s.$$

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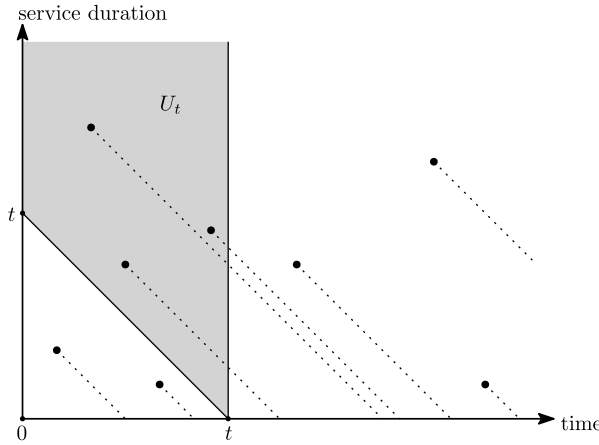


FIGURE 1. A Poisson representation of $M/G/\infty$. In this example, there are *exactly* three customers at time t .

In particular, if

$$\int_0^\infty (\mathbb{E}(S \wedge t))^k \exp(-\lambda \mathbb{E}(S \wedge t)) dt < \infty \quad \text{for all } k \geq 0, \tag{2}$$

then the system is transient; if

$$\int_0^\infty \exp(-\lambda \mathbb{E}(S \wedge t)) dt = \infty,$$

then the system is recurrent.

Before proving this result, we make the following remark. Let us define $M(t)$ to be the maximal remaining service time of the customers that are present at time t . This is a so-called *extremal shot noise process*; see [1] and references therein. It is not difficult to obtain that transience of $M(\cdot)$ is the same as transience of state 0 in $M/G/\infty$; then Theorem 2.5 of [1] provides a criterion for the transience of $M(\cdot)$ (and therefore for the transience of state 0 in our situation).

Proof of Theorem 1. We start with a simple observation: for any $j \geq 0$, $\{\liminf Y_t = j\}$ is a tail event, so it has probability 0 or 1. This implies that $\liminf Y_t$ is a.s. a constant (which may be equal to $+\infty$).

We use the following representation of the process (see Figure 1): consider a Poisson process in \mathbb{R}_+^2 , with the intensity measure $\lambda dt \times dF_S(u)$, where $F_S(u) = \mathbb{P}[S \leq u]$ is the distribution function of S . Then, a point (t, u) of this Poisson process is interpreted in the following way: a customer arrived at time t and the duration of its service will be u . Now, draw a (dotted) line in the southeast direction from each point, as shown in the picture; as long as this line stays in \mathbb{R}_+^2 , the corresponding customer is present in the system. If we draw a vertical line from $(t, 0)$ in the upwards direction, then the number of dotted lines it intersects is equal to Y_t .

Next, for $k \in \mathbb{Z}_+$ let

$$T_k := \{t \geq 0 : Y_t = k\}$$

denote the set of times when the system has exactly k customers, and let

$$U_t = \{(s, u) \in \mathbb{R}_+^2 : s \in [0, t], u \geq t - s\}.$$

We note that Y_t equals the number of points in U_t , which has Poisson distribution with mean

$$\int_{U_t} \lambda \, dt \, dF_S(u) = \lambda \mathbb{E}(S \wedge t).$$

Therefore, by Fubini’s theorem, we have (here $|A|$ stands for the Lebesgue measure of $A \subset \mathbb{R}$)

$$\mathbb{E}|T_k| = \mathbb{E} \int_0^\infty \mathbf{1}\{Y_t = k\} \, dt = \frac{\lambda^k}{k!} \int_0^\infty (\mathbb{E}(S \wedge t))^k \exp(-\lambda \mathbb{E}(S \wedge t)) \, dt. \tag{3}$$

Now, assume that $\mathbb{E}|T_k| < \infty$ for some $k \geq 0$; it automatically implies that $\mathbb{E}|T_\ell| < \infty$ for $0 \leq \ell \leq k$. This means that $|T_0|, \dots, |T_k|$ are a.s. finite, and let us show that T_0, \dots, T_k have to be a.s. bounded (this is a small technical issue that we have to resolve because we are considering continuous time). Probably the cleanest way to see this is as follows. First notice that, in fact, T_0 is a union of intervals of random i.i.d. (with $\text{Exp}(\lambda)$ distribution) lengths, because each time the system becomes empty it will remain so until the arrival of the next customer. Therefore, $|T_0| < \infty$ clearly means that $\sup T_0 \leq K_0$ for some (random) K_0 . Now, *after* K_0 there are no longer any $1 \rightarrow 0$ transitions, so the remaining part of T_1 again becomes a union of such intervals, meaning that it should be bounded as well; we then repeat this reasoning a suitable number of times to finally obtain that T_k must be a.s. bounded. This implies that $\liminf_{t \rightarrow \infty} Y_t \geq k_0$ a.s.

Next, assume that $\{0, \dots, k\}$ is a *transient set*, in the sense that $\liminf_{t \rightarrow \infty} Y_t \geq k + 1$ a.s.; let us show that this implies that $\mathbb{E}|T_k| < \infty$. Indeed, first we can choose a sufficiently large $h > 0$ in such a way that

$$\mathbb{P}[Y_t \geq k + 1 \text{ for all } t \geq h] \geq \frac{1}{2}.$$

Define a stopping time $\tau = \inf\{t \geq h : Y_t \leq k\}$ (again, with the convention $\inf \emptyset = +\infty$). Then, a crucial observation is that what one sees after τ is a superposition of two *independent* systems: one is formed by those customers (with their remaining lifetimes) present at τ , and the other is a copy of the original system. Then, a simple coin-tossing argument, together with the fact that an initially non-empty system (i.e. with some customers being served, with any assumptions on their remaining service times) dominates an initially empty system, shows that $|T_k|$ (in fact, $|T_0| + \dots + |T_k|$) is dominated by $h \times \text{Geom}_0(\frac{1}{2})$ random variable and therefore has a finite expectation. It means that we have $\liminf_{t \rightarrow \infty} Y_t \leq k_0$ a.s. (because otherwise, in the situation when $k_0 < \infty$, we would have $\mathbb{E}|T_{k_0}| < \infty$, which, by definition, is not the case). This concludes the proof of Theorem 1. □

Regarding this result, we may observe that in most situations one would have $k_0 = 0$ or $+\infty$; this is because convergence of such integrals is usually determined by what is in the exponent. Still, it is not difficult to construct ‘strange examples’ with $0 < k_0 < \infty$, i.e. where the process will visit $\{0, \dots, k_0 - 1\}$ only finitely many times, but will hit every $k \geq k_0$ infinitely often a.s. (a behaviour one cannot have with irreducible Markov chains). For instance, let $\lambda = 1$ and fix $b > 0$; next, consider a service time distribution such that

$$1 - F_S(u) = \frac{1}{u} + \frac{b}{u \ln u}$$

for large enough u . Then it is elementary to obtain that $\mathbb{E}(S \wedge t) = \ln t + b \ln \ln t + O(1)$ and the integrals in (1) diverge whenever $k \geq b - 1$, meaning that $k_0 = \lceil b \rceil - 1$.

Now, in the situation when (2) holds and Y is transient, it may also be useful to be able to say something about the speed of convergence of Y_t to infinity. We do not intend to enter deeply into this question here, but only prove a particular result needed for future reference. Namely, in [2] we work with a different model which in some sense dominates $M/G/\infty$. So, we will now give a lower bound on the growth of Y_t ; more specifically, we will show that under certain conditions Y_t will eventually be at least a constant fraction of its expected value. For $q \in (0, 1)$, let us define $\gamma_q = 1 - q - q \ln q^{-1} > 0$.

Theorem 2. Fix $q \in (0, 1)$ and assume that

$$\int_0^\infty \exp(-\gamma_q \lambda \mathbb{E}(S \wedge t)) dt < \infty. \quad (4)$$

Then

$$\mathbb{P}[Y_t \geq q\lambda \mathbb{E}(S \wedge t) \text{ for all large enough } t] = 1. \quad (5)$$

Proof. Let

$$H_q = \{t \geq 0 : Y_t < q\lambda \mathbb{E}(S \wedge t)\};$$

our goal is to show that H_q is a.s. bounded in the case when (4) holds. We recall a standard (Chernoff) tail bound: if X is Poisson(μ) and $q \in (0, 1)$, then

$$\mathbb{P}[X \leq q\mu] \leq \exp(-(q\mu \ln q + \mu - q\mu)) = \exp(-\gamma_q \mu). \quad (6)$$

Then, analogously to (3), we obtain from (6) that

$$\mathbb{E}|H_q| \leq \int_0^\infty \exp(-\gamma_q \lambda \mathbb{E}(S \wedge t)) dt; \quad (7)$$

so, by (4), we have $\mathbb{E}|H_q| < \infty$, meaning that $|H_q| < \infty$ a.s. To see that this has to imply that H_q is a.s. bounded, analogously to the proof of Theorem 1, one can reason in the following way. If $t \in H_q$, then $s \in H_q$ for all $s \in (t, A_t)$, where A_t is the first moment after t when a customer arrives in the system. This implies that the lengths of the intervals that constitute H_q dominate a sequence of i.i.d. random variables with $\text{Exp}(\lambda)$ distribution; in its turn, this clearly implies that if $|H_q|$ is finite then it has to be bounded. \square

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