

## ON ERLANG'S FORMULA<sup>1</sup>

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**1. Introduction.** The following mathematical model of telephone traffic has some importance in designing telephone exchanges.

In the time interval  $(0, \infty)$  calls are arriving at a telephone exchange in accordance with a Poisson process of density  $\lambda$ , that is, if  $\tau_1, \tau_2, \dots, \tau_n, \dots$  denote the arrival times, then  $\tau_n - \tau_{n-1}$  ( $n = 1, 2, \dots; \tau_0 = 0$ ) are mutually independent random variables having a common distribution function

$$(1) \quad \begin{aligned} F(x) &= 1 - e^{-\lambda x} & \text{if } x \geq 0, \\ &= 0 & \text{if } x < 0. \end{aligned}$$

There are  $m$  available lines. If an arriving call finds a free line, then a connection is realized without delay. If every line is busy when a call arrives, the call is lost. The holding times are mutually independent, positive random variables having a common distribution function  $H(x)$ , and a finite expectation

$$(2) \quad \alpha = \int_0^\infty x dH(x).$$

The holding times are also independent of the arrival times and the initial state. The initial state is given by the number of busy lines at time  $t = 0$  and by the remaining lengths of the holding times in progress at time  $t = 0$ .

If we choose the initial distribution in such a way that the process becomes stationary, then the probability that at time  $t$  the number of busy lines is  $k$  is given by Erlang's formula,

$$(3) \quad P_k = [(\lambda\alpha)^k / k!] [\sum_{j=0}^m (\lambda\alpha)^j / j!]^{-1}$$

for  $k = 0, 1, \dots, m$  and all  $t \geq 0$ .

This formula has an interesting history. In 1917 A. K. Erlang [1] deduced formula (3) for the case when the holding times are constant  $\alpha$ . While Erlang's result is correct, his proof is not complete. He has made use of a property of the process which is far from evident. Erlang noted also that if the holding times have an exponential distribution, that is,

$$(4) \quad \begin{aligned} H(x) &= 1 - e^{-x/\alpha} & \text{for } x \geq 0, \\ &= 0 & \text{for } x < 0, \end{aligned}$$

then (3) is valid. If  $H(x)$  is an exponential distribution function, then Erlang's proof is acceptable.

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Erlang's investigations automatically raised the problem of whether (3) remains valid also for an arbitrary  $H(x)$ . The answer is affirmative and this was established first in 1927 by A. E. Vulot [11], and subsequently by F. Pollaczek [8], C. Palm [7], L. Kosten [5], K. Lundkvist [6], A. Y. Khinchin [4] and B. A. Sevastyanov [9]. For other appropriate references see R. Syski [10] pp. 271–278.

There is an essential difference between the cases of an exponential  $H(x)$  and an arbitrary  $H(x)$ . If we denote by  $\nu(t)$  the number of busy lines at time  $t$ , then  $\{\nu(t), 0 \leq t < \infty\}$  is a Markov process if and only if  $H(x)$  is an exponential distribution function. If  $H(x)$  is given by (4), then it is easy to show that

$$(5) \quad \lim_{t \rightarrow \infty} \mathbf{P}\{\nu(t) = k\} = P_k$$

exists, is independent of the distribution of  $\nu(0)$ , and  $P_k (k = 0, 1, \dots, m)$  is given by (3). If we assume that  $\mathbf{P}\{\nu(0) = k\} = P_k (k = 0, 1, \dots, m)$ , then  $\{\nu(t), 0 \leq t < \infty\}$  becomes a stationary process for which  $\mathbf{P}\{\nu(t) = k\} = P_k$  for all  $t \geq 0$ . That is for a stationary process,  $P_k (k = 0, 1, \dots, m)$  is the probability that at time  $t$  the number of busy lines is  $k$ .

In the general case,  $\{\nu(t), 0 \leq t < \infty\}$  is not a Markov process. However, if we introduce auxiliary variables we can achieve that the process becomes Markovian. If  $\nu(t) = k (k = 1, 2, \dots, m)$ , then let  $\chi_1(t), \chi_2(t), \dots, \chi_k(t)$  be a random permutation of the remaining lengths of the  $k$  holding times in progress at time  $t$ . We suppose that all the  $k!$  permutations are equally probable. Then the process  $\{\nu(t), \chi_1(t), \dots, \chi_{\nu(t)}(t); 0 \leq t < \infty\}$  is a Markov process. It can be shown that

$$(6) \quad \lim_{t \rightarrow \infty} \mathbf{P}\{\nu(t) = k, \chi_1(t) \leq x_1, \dots, \chi_k(t) \leq x_k\} = P_k \prod_{i=1}^k H^*(x_i)$$

for  $k = 0, 1, \dots, m$  and  $x_1 \geq 0, x_2 \geq 0, \dots, x_m \geq 0$ , where

$$(7) \quad \begin{aligned} H^*(x) &= \alpha^{-1} \int_0^x [1 - H(u)] du & \text{if } x \geq 0, \\ &= 0 & \text{if } x < 0, \end{aligned}$$

and  $P_k$  is defined by (3). The limit (6) is independent of the distribution of  $(\nu(0), \chi_1(0), \dots, \chi_{\nu(0)}(0))$ . If we suppose that the distribution of  $(\nu(0), \chi_1(0), \dots, \chi_{\nu(0)}(0))$  agrees with the limiting distribution (6), then  $\{\nu(t), \chi_1(t), \dots, \chi_{\nu(t)}(t)\}$  becomes a stationary process for which  $\nu(t)$  has the same distribution for all  $t \geq 0$ , namely,  $\mathbf{P}\{\nu(t) = k\} = P_k$  for  $k = 0, 1, \dots, m$  and all  $t \geq 0$ .

We note that in some of the papers mentioned above only  $\{P_k\}$  has been found. F. Pollaczek [8], K. Lundkvist [6] and A. Y. Khinchin [4] proved only Erlang's formula (3). C. Palm [7], L. Kosten [5], and B. A. Sevastyanov [9] proved also (6). Actually they interpreted  $\chi_1(t), \dots, \chi_{\nu(t)}(t)$  as the lengths of the past durations of the holding times in progress at time  $t$ . The interpretation used in this paper has also been used by S. Erlander [2].

As far as the proof of (6) is concerned, it consists of two parts. First, the ergodicity of the process should be proved. This is based on the use of rather deep theorems. Second, it requires the solution of a system of integral equations and the proof of the existence and the uniqueness of the solution. Most of the proofs

mentioned above are intuitive and contain heuristic reasoning. From the mathematical point of view the proof of B. A. Sevastyanov [9] is the most satisfactory.

There is one thing that is worth special mention. The probability (3) is usually interpreted as the probability that an arriving call finds  $k$  lines busy in a stationary process. Let us denote by  $\nu_n$  the number of lines that the  $n$ th arriving call finds busy. If we suppose that the process  $\{\nu(t), \chi_1(t), \dots, \chi_{\nu(t)}(t)\}$  is stationary, then the distribution of  $\nu_n$  depends on  $n$ , that is,  $\{\nu_n\}$  is not a stationary sequence. If we want to be exact, the latter interpretation of probability (3) fails. However, in most of the applications, our interest is in finding the probability that an arriving call finds  $k$  lines busy, or, in particular, all the  $m$  lines busy.

In what follows we shall study the limiting distribution of  $\nu_n$  as  $n \rightarrow \infty$ . By choosing a suitable initial distribution we shall define a process for which  $\nu_n$  has the same distribution for all  $n = 1, 2, \dots$ , that is, for which  $\{\nu_n\}$  is a stationary sequence.

It turns out that if we investigate the distribution of  $\nu_n$  ( $n = 1, 2, \dots$ ) instead of  $\nu(t)$  ( $0 \leq t < \infty$ ), then everything becomes very simple. To prove the ergodicity of  $\{\nu_n\}$  we need to refer only to an elementary theorem of recurrent events and to find the limiting distribution of  $\nu_n$  as  $n \rightarrow \infty$  we need to use only integration by parts.

It is interesting to note that by this shift of attention from the time-dependent behavior of the process to the arrival-dependent behavior, all mathematical difficulties disappear.

**2. General holding times.** Consider the mathematical model of telephone traffic formulated at the beginning of the Introduction. Let  $\nu_n$  be the number of lines busy immediately before the arrival of the  $n$ th call. If  $\nu_n = k$  ( $k = 1, 2, \dots, m$ ), then let  $(\chi_{n1}, \dots, \chi_{nk})$  be a random permutation of the remaining lengths of the  $k$  holding times in progress at time  $\tau_n$ . We suppose that all the  $k!$  permutations are equally probable. Now we shall prove that the following theorem holds.

**THEOREM.** *We have*

$$(8) \quad \lim_{n \rightarrow \infty} \mathbf{P}\{\nu_n = k, \chi_{n1} > x_1, \dots, \chi_{nk} > x_k\} = P_k \prod_{i=1}^k [1 - H^*(x_i)]$$

for  $k = 0, 1, \dots$  and  $x_1 \geq 0, x_2 \geq 0, \dots, x_m \geq 0$ , where

$$(9) \quad P_k = [(\lambda\alpha)^k / k!] [\sum_{j=0}^m (\lambda\alpha)^j / (j!)]^{-1}$$

for  $k = 0, 1, \dots, m$  and

$$(10) \quad H^*(x) = \alpha^{-1} \int_0^x [1 - H(u)] du \quad \text{for } x \geq 0, \\ = 0 \quad \text{for } x < 0.$$

*The limit (8) is independent of the initial state.*

**PROOF.** The vector sequence  $\xi_n = (\nu_n, \chi_{n1}, \dots, \chi_{n\nu_n})$ , ( $n = 1, 2, \dots$ ) is a discrete parameter Markov process. First we shall show that the distribution de-

finned by (8) is a stationary distribution of the process, that is, if we assume that  $\xi_n$  has the distribution (8), then it follows that  $\xi_{n+1}$  has the same distribution. (8) is indeed a stationary distribution, if the following equations hold:

$$(11) \quad P_0 = P_0 \int_0^\infty H(y)e^{-\lambda y} dy + \cdots + P_{m-1} \int_0^\infty [H^*(y)]^{m-1} H(y)e^{-\lambda y} dy \\ + P_m \int_0^\infty [H^*(y)]^m e^{-\lambda y} dy,$$

and for  $k = 1, 2, \dots, m$  and  $x_1 \geq 0, x_2 \geq 0, \dots, x_m \geq 0$ ,

$$(12) \quad P_k [1 - H^*(x_1)] \cdots [1 - H^*(x_k)] \\ = k^{-1} P_{k-1} \int_0^\infty [1 - H^*(x_1 + y)] \cdots [1 - H^*(x_k + y)] \\ \cdot \left\{ \sum_{i=1}^k [1 - H(x_i + y)] [1 - H^*(x_i + y)]^{-1} \right\} e^{-\lambda y} dy \\ + \cdots + P_{m-1} \int_0^\infty [1 - H^*(x_1 + y)] \cdots [1 - H^*(x_k + y)] \\ \cdot \left\{ \binom{m-1}{k} [H^*(y)]^{m-1-k} H(y) + k^{-1} \binom{m-1}{k-1} [H^*(y)]^{m-k} \right. \\ \left. \cdot \sum_{i=1}^k [1 - H(x_i + y)] [1 - H^*(x_i + y)]^{-1} \right\} e^{-\lambda y} dy \\ + P_m \binom{m}{k} \int_0^\infty [1 - H^*(x_1 + y)] \cdots [1 - H^*(x_k + y)] [H^*(y)]^{m-k} e^{-\lambda y} dy$$

where  $P_k (k = 0, 1, \dots, m)$  is defined by (9).

In obtaining the right hand side of (11) we took into consideration that the event that the  $(n+1)$ st call finds no busy line can occur in the following ways:  $\tau_{n+1} - \tau_n = y$  where  $0 \leq y < \infty, \nu_n = r$  where  $r = 0, 1, \dots, m$  and the remaining lengths of the current holding times at  $\tau_n$  and the length of the holding time beginning at  $\tau_n$  (if  $r < m$ ) are all  $\leq y$ .

In obtaining the right hand side of (12) we took into consideration that the event that the  $(n+1)$ st call finds  $k$  lines busy and the remaining lengths of the current holding times at time  $\tau_{n+1}$  are greater than  $x_1, x_2, \dots, x_k$  respectively, can occur in the following ways:  $\tau_{n+1} - \tau_n = y$  where  $0 \leq y < \infty, \nu_n = r$  where  $r = k-1, k, \dots, m$  and among the remaining lengths of the current holding times at time  $\tau_n$  and the length of the holding time beginning at  $\tau_n$  (if  $r < m$ )  $k$  are greater than  $x_1 + y, \dots, x_k + y$  respectively and all the others are  $\leq y$ .

Now we have to check whether the left hand sides of (11) and (12) are equal to the corresponding right hand sides or not. It is easy to show that both (11) and (12) hold for any  $m = 1, 2, \dots$  and  $x_1 \geq 0, \dots, x_m \geq 0$ . It will be convenient to use the following abbreviations for fixed  $x_1 \geq 0, \dots, x_m \geq 0$ ,

$$(13) \quad A_j(y) = [1 - H^*(x_1 + y)] \cdots [1 - H^*(x_j + y)]$$

if  $j = 1, 2, \dots, m$ , and

$$(14) \quad B_j(y) = [H^*(y)]^j$$

for  $j = 0, 1, 2, \dots$ . Then we have

$$(15) \quad dA_j(y)/dy = -A_j(y)\alpha^{-1} \sum_{i=1}^j [1 - H(x_i + y)] [1 - H^*(x_i + y)]^{-1},$$

and

$$(16) \quad dB_j(y)/dy = j[1 - H(y)]\alpha^{-1} B_{j-1}(y).$$

By using the above notation, (11) and (12) can be written in the following equivalent forms:

$$(17) \quad 1 = \int_0^\infty H(y)e^{-\lambda y} \lambda dy + \dots + (\lambda\alpha)^{m-1}[(m-1)!]^{-1} \cdot \int_0^\infty B_{m-1}(y)H(y)e^{-\lambda y} \lambda dy + (\lambda\alpha)^m(m!)^{-1} \int_0^\infty B_m(y)e^{-\lambda y} \lambda dy$$

and for  $k = 1, 2, \dots, m$ ,

$$(18) \quad \begin{aligned} & (\lambda\alpha)^k(k!)^{-1}A_k(0) \\ &= -(\lambda\alpha)^k(k!)^{-1} \int_0^\infty (dA_k(y)/dy)e^{-\lambda y} dy + \dots \\ &+ (\lambda\alpha)^{m-1}[(m-1)!]^{-1} \int_0^\infty \binom{m-1}{k} A_k(y)B_{m-k-1}(y)H(y) \\ &- \alpha k^{-1} \binom{m-1}{k-1} (dA_k(y)/dy)B_{m-k}(y) e^{-\lambda y} \lambda dy \\ &+ (\lambda\alpha)^m(m!)^{-1} \binom{m}{k} \int_0^\infty A_k(y)B_{m-k}(y)e^{-\lambda y} \lambda dy. \end{aligned}$$

First, we prove (17). If  $m = 1$ , then (17) reduces to

$$(19) \quad \lambda\alpha \int_0^\infty B_1(y)e^{-\lambda y} \lambda dy = \int_0^\infty [1 - H(y)]e^{-\lambda y} \lambda dy = \alpha \int_0^\infty (dB_1(y)/dy)e^{-\lambda y} \lambda dy,$$

which can be seen to be true by integrating by parts. Now we shall prove by mathematical induction that (17) is true for all  $m = 1, 2, \dots$ . Suppose that (17) is true for  $m$  ( $m = 1, 2, \dots$ ). The difference between the right hand side of (17) for  $m + 1$  and for  $m$  is

$$(20) \quad (\lambda\alpha)^{m+1}[(m+1)!]^{-1}[\int_0^\infty B_{m+1}(y)e^{-\lambda y} \lambda dy - \int_0^\infty (dB_{m+1}(y)/dy)e^{-\lambda y} dy] = 0,$$

which follows by integrating by parts. Thus (17) holds also for  $m + 1$ . Accordingly (17) is true for all  $m = 1, 2, \dots$ .

Second, we prove (18). If  $m = k$ , then (18) reduces to

$$(21) \quad \begin{aligned} & (\lambda\alpha)^k(k!)^{-1}A_k(0) \\ &= (\lambda\alpha)^k(k!)^{-1}[\int_0^\infty A_k(y)e^{-\lambda y} \lambda dy - \int_0^\infty (dA_k(y)/dy)e^{-\lambda y} dy] \end{aligned}$$

which is evidently true. Now we shall prove by mathematical induction that (18) is true for all  $m = k, k + 1, \dots$ . Suppose that (18) is true for  $m$  ( $m = k, k + 1, \dots$ ). The difference between the right hand side of (18) for  $m + 1$  and  $m$  is

$$(22) \quad \begin{aligned} & (\lambda\alpha)^{m+1}[(m+1)!]^{-1}[\int_0^\infty A_k(y)B_{m+1-k}(y)e^{-\lambda y} \lambda dy \\ &- \int_0^\infty (dA_k(y)/dy)B_{m+1-k}(y)e^{-\lambda y} dy \\ &- \int_0^\infty A_k(y)(dB_{m+1-k}(y)/dy)e^{-\lambda y} dy] = 0 \end{aligned}$$

which follows again by integrating by parts. Thus (18) holds also for  $m + 1$ . Accordingly (18) is true for all  $m = k, k + 1, \dots$ .

We can conclude that  $\xi_n = (\nu_n, \chi_{n1}, \dots, \chi_{nn})$  ( $n = 1, 2, \dots$ ) has a stationary distribution defined by (8). If we suppose that  $\xi_1$  has the distribution defined by (8), then every  $\xi_n$  ( $n = 1, 2, \dots$ ) will have the same distribution as  $\xi_1$ .

We observe that the event that an arriving call finds all the  $m$  lines free is a recurrent event. This event is aperiodic and it occurs at the  $n$ th arrival if  $\nu_n = 0$ . Consequently,

$$(23) \quad \lim_{n \rightarrow \infty} \mathbf{P}\{\nu_n = 0\}$$

exists, and is independent of the initial state. If, in particular,  $\{\xi_n\}$  is the stationary sequence considered above, then we have seen that  $\mathbf{P}\{\nu_n = 0\} = P_0$  for all  $n = 1, 2, \dots$ . Thus it follows that the limit (23) is necessarily  $P_0$ . Accordingly, the recurrent event is persistent, and the mean recurrence time is  $1/P_0$ . (See W. Feller [3] Chapter XIII.) This fact implies that the limiting distribution of  $\xi_n$  is independent of the initial distribution. For any time when an arriving call finds all the  $m$  lines free, the future stochastic behavior of the process is the same independently of the past. In a particular case we found the limiting distribution of  $\xi_n$ , namely for the stationary process defined above. Hence, regardless of the initial distribution,  $\xi_n$  has the limiting distribution (8). This completes the proof of the Theorem.

Evidently (8) is the unique stationary distribution of  $\{\xi_n\}$ . We note that  $\{\xi_n\}$  becomes a stationary sequence if we suppose that the distribution of  $(\nu(0), \chi_1(0), \dots, \chi_{\nu(0)}(0))$  is as follows:  $\mathbf{P}\{\nu(0) = 0\} = 0$ ,

$$(24) \quad \begin{aligned} \mathbf{P}\{\nu(0) = k, \chi_1(0) > x_1, \dots, \chi_k(0) > x_k\} \\ = k^{-1} P_{k-1} [1 - H^*(x_1)] \cdots [1 - H^*(x_k)] \\ \cdot \sum_{i=1}^k [1 - H(x_i)] [1 - H^*(x_i)]^{-1} \end{aligned}$$

for  $k = 1, 2, \dots, m - 1$  and

$$(25) \quad \begin{aligned} \mathbf{P}\{\nu(0) = m, \chi_1(0) > x_1, \dots, \chi_m(0) > x_m\} \\ = [1 - H^*(x_1)] \cdots [1 - H^*(x_m)] \\ \cdot \{m^{-1} P_{m-1} \sum_{i=1}^m [1 - H(x_i)] [1 - H^*(x_i)]^{-1} + P_m\}. \end{aligned}$$

In this case the probability that the  $n$ th arriving call finds  $k$  lines busy is given by  $P_k$  ( $k = 0, 1, \dots, m$ ) for every  $n = 1, 2, \dots$ .

**3. Exponential holding times.** If we suppose that the holding times have an exponential distribution

$$(26) \quad \begin{aligned} H(x) &= 1 - e^{-\mu x} & \text{if } x \geq 0, \\ &= 0 & \text{if } x < 0, \end{aligned}$$

and  $\nu_n$  denotes the number of busy lines immediately before the arrival of the  $n$ th call, then we can see easily that  $\{\nu_n\}$  is a Markov chain. This Markov chain is homogeneous, irreducible and aperiodic. The transition probabilities  $\mathbf{P}\{\nu_{n+1} = k \mid \nu_n = j\} = p_{jk}$  can be expressed as follows:

$$(27) \quad p_{jk} = \binom{j+1}{k} \int_0^\infty (1 - e^{-\mu x})^{j+1-k} e^{-k\mu x} e^{-\lambda x} \lambda dx$$

for  $j = 0, 1, \dots, m - 1$  and  $p_{m,k} = p_{m-1,k}$ .

In this case the limiting distribution  $\lim_{n \rightarrow \infty} \mathbf{P}\{\nu_n = k\} = P_k$  ( $k = 0, 1, \dots, m$ ) exists, is independent of the initial distribution and  $\{P_k\}$  is the only stationary distribution. Now we are going to find  $\{P_k\}$ . We note that if we know the binomial moments

$$(28) \quad B_r = \sum_{k=r}^m \binom{k}{r} P_k$$

for  $r = 0, 1, \dots, m$ , then  $\{P_k\}$  can be obtained by

$$(29) \quad P_k = \sum_{r=k}^m (-1)^{r-k} \binom{r}{k} B_r \quad (k = 0, 1, \dots, m).$$

For the determination of the binomial moments  $B_0, B_1, \dots, B_m$  we can deduce a recurrence formula. If  $\nu_n = j$  and  $\tau_{n+1} - \tau_n = x$ , then  $\nu_{n+1}$  has a Bernoulli distribution  $B(j+1, e^{-\mu x})$  for  $j = 0, 1, \dots, m-1$  and  $B(m, e^{-\mu x})$  for  $j = m$ . Thus

$$(30) \quad \begin{aligned} & \mathbf{E}\{\binom{\nu_{n+1}}{r} \mid \nu_n = j\} \\ &= \binom{j+1}{r} \int_0^\infty e^{-r\mu x - \lambda x} \lambda dx = \binom{j+1}{r} \lambda (\lambda + r\mu)^{-1} \quad \text{if } j = 0, 1, \dots, m-1, \\ &= \binom{m}{r} \int_0^\infty e^{-r\mu x - \lambda x} \lambda dx = \binom{m}{r} \lambda (\lambda + r\mu)^{-1} \quad \text{if } j = m. \end{aligned}$$

If  $\{\nu_n\}$  has a stationary distribution  $\{P_k\}$ , then  $\mathbf{E}\{\binom{\nu_n}{r}\} = B_r$  for all  $n = 1, 2, \dots$ . Then by (30) we obtain that

$$(31) \quad (\lambda + r\mu)B_r = \lambda B_r + \lambda B_{r-1} - \lambda \binom{m}{r-1} B_m,$$

or, in a simpler form,

$$(32) \quad r\mu B_r = \lambda B_{r-1} - \lambda \binom{m}{r-1} B_m$$

for  $r = 1, 2, \dots, m$ .

If we put (32) into (29) and we use (28), then we can write that

$$(33) \quad P_k = \lambda (\mu k)^{-1} P_{k-1}$$

for  $k = 1, 2, \dots, m$ , whence it follows that

$$(34) \quad P_k = [(\lambda/\mu)^k / k!] [\sum_{j=0}^m (\lambda/\mu)^j (j!)^{-1}]^{-1}$$

for  $k = 0, 1, \dots, m$ . This is in agreement with (9) because now  $\alpha = 1/\mu$  is the expectation of the holding times.

We note that if we assume that  $\mathbf{P}\{\nu(0) = 0\} = 0$ ,  $\mathbf{P}\{\nu(0) = k\} = P_{k-1}$  for  $k = 1, 2, \dots, m-1$  and  $\mathbf{P}\{\nu(0) = m\} = P_{m-1} + P_m$ , then  $\{\nu_n\}$  becomes a stationary Markov chain for which  $\mathbf{P}\{\nu_n = k\} = P_k$  ( $k = 0, 1, \dots, m$ ) for all  $n = 1, 2, \dots$ . In this case  $\{\nu(t), 0 \leq t < \infty\}$  is not a stationary Markov process. Conversely, if  $\{\nu(t), 0 \leq t < \infty\}$  is a stationary Markov process, then  $\{\nu_n\}$  is not a stationary sequence.

**4. Infinitely many lines.** Suppose that in the telephone exchange every arriving call realizes a connection without delay, that is, there is no lost call. In this case  $m$ , the number of available lines, is infinite. Denote by  $\nu_n$  the number of busy lines immediately before the arrival of the  $n$ th call. The distribution of  $\nu_n$

can easily be found. If at time  $t = 0$  there is no busy line, then we have

$$(35) \quad \mathbf{P}\{\nu_{n+1} = k\} \\ = \binom{n}{k} \lambda^{n+1} (n!)^{-1} \int_0^\infty e^{-\lambda x} \left[ \int_0^x H(u) du \right]^{n-k} \left\{ \int_0^x [1 - H(u)] du \right\}^k dx.$$

In case of any other initial state, (35) needs only obvious modifications. Regardless of the initial distribution we have the following limit:

$$(36) \quad \lim_{n \rightarrow \infty} \mathbf{P}\{\nu_n = k\} = e^{-\lambda \alpha} (\lambda \alpha)^k (k!)^{-1}$$

for  $k = 0, 1, 2, \dots$ .

If  $\nu_n = k$ , then let  $\chi_{n1}, \dots, \chi_{nk}$  be a random permutation of the remaining lengths of the  $k$  holding times in progress at the arrival of the  $n$ th call. Suppose that all the  $k!$  permutations are equally probable. If we suppose that at time  $t = 0$  there is no busy line, then for  $x_1 \geq 0, \dots, x_k \geq 0$  we have

$$(37) \quad \mathbf{P}\{\nu_{n+1} = k, \chi_{n+1,1} > x_1, \dots, \chi_{n+1,k} > x_k\} \\ = \binom{n}{k} \lambda^{n+1} (n!)^{-1} \int_0^\infty e^{-\lambda x} \left[ \int_0^x H(u) du \right]^{n-k} \left\{ \prod_{i=1}^k \int_0^x [1 - H(u + x_i)] du \right\} dx,$$

whence

$$(38) \quad \lim_{n \rightarrow \infty} \mathbf{P}\{\nu_n = k, \chi_{n1} > x_1, \dots, \chi_{nk} > x_k\} \\ = e^{-\lambda \alpha} (\lambda \alpha)^k (k!)^{-1} \prod_{i=1}^k [1 - H^*(x_i)],$$

where  $H^*(x)$  is defined by (10). If we consider an arbitrary initial distribution, then (37) needs only obvious modifications and (38) remains valid regardless of the initial distribution.

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