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ON THE WAITING TIME DISTRIBUTION OF  
BULK QUEUES

by

John Dagsvik

## Abstract

The waiting time process of the  $n$ -th arriving group is considered for the general bulk queueing model  $GI^X/G^Y/1$ .

A generalisation of Lindley's waiting time equation is established.

By a generalisation of Kingman's method [3], this equation is solved for the models  $GI^X/E_k^Y/1$  and  $E_k^X/G^Y/1$ .

When the service time is Erlang distributed  $E_k$ , the results are applied to the case where the service- and the arrival groups are of constant size.

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Key words: Wendel projection, Group Waiting time, Restbatch, Waiting time equation, Erlang distributions, Hyperexponential distribution, Stationary distributions.

## Contents

	page
1. Introduction	1
2. The algebraic formalism	1
3. Group waiting time	3
4. The model $GI^X/E_k^Y/1$	12
5. Special cases	16
6. The model $E_k^X/G^Y/1$	19
7. The stationary solution	21

## 1. Introduction

In the present paper we shall assume that customers arrive in groups  $C_n$ ,  $n = 0, 1, 2, \dots$ . The group size is a stochastic variable  $X$ , with probability distribution  $f(\cdot)$ . The inter-arrival intervals  $A_n$ ,  $n = 0, 1, \dots$  are independent and have the same distribution  $a(\cdot)$ . The service mechanism is described as follows: At the end of a service period the server accepts  $Y$  customers from the waiting line, or a smaller number if the line is shorter.  $Y$  is called the service group capacity. The length of the service time  $B$ , has the distribution  $b(\cdot)$ . We shall assume the existence of two integers  $m, l$  such that  $X \leq l$ ,  $Y \leq m$ .

The most general works on bulk queues seems to be those of Keilson [2], Cohen [1], Le Gall [5], Lambotte and Teghem [4]. They obtain the distribution of the queue length from which the waiting time distribution is derived. However, there exists no such results for general distributions  $a(\cdot)$ ,  $b(\cdot)$ ,  $f(\cdot)$  and  $g(\cdot)$ . Earlier works are restricted to the case where  $a(\cdot)$  or  $b(\cdot)$  are the exponential distribution. Even if  $b(\cdot)$  is exponential the analysis are only limited to bulk service models (Cohen [1], Le Gall [5]).

## 2. The algebraic formalism.

Let  $\Omega_n$  denote the set of  $n \times n$  matrices whose components are finite complex measures on the Borel subsets of the real line. According to Kingman [3] the product of two measures is defined as their convolution. An operator  $T : \Omega_1 \rightarrow \Omega_1$  is defined by

$$(2.1) \quad (Tv)(E) = v(E \cap R^+) + v(-R^+) \epsilon(E) , \quad v \in \Omega_1 ,$$

where  $\epsilon$  is the measure

$$\epsilon(E) = \begin{cases} 1 & \text{if } 0 \in E \\ 0 & \text{if } 0 \notin E \end{cases}$$

and  $R^+ = (0, \infty)$ .

This operator has the property that if  $X$  is a random variable with distribution

$$v(E) = \Pr\{X \in E\} ,$$

then

$$(Tv)(E) = \Pr\{X^+ \in E\} .$$

Kingman shows that  $\Omega_1$  is an commutative algebra over the complex field  $\mathbb{C}$  with identity  $\epsilon$ . Furthermore, he shows that the image  $\Omega_1^+$  and the kernel  $\Omega_1^-$  of  $T$  are both disjoint subalgebras of  $\Omega_1$ .  $T$  is extended to  $\Omega_n$  by  $T\{v_{ij}\} = \{Tv_{ij}\}$ ,  $\{v_{ij}\} \in \Omega_n$  and multiplication in  $\Omega_n$  is defined in the obvious way. With  $I_n \epsilon$  as identity it is easy to verify that  $\Omega_n$  has the same properties as  $\Omega_1$  except that  $\Omega_n$  is no longer commutative. The norm on  $\Omega_n$  is defined by

$$\|v\| = \max_j \sum_i |dv_{ij}| , \quad v = \{v_{ij}\} .$$

The set whose elements are Fourier-Stieltjes transform of elements from  $\Omega_1$  we denote  $\hat{\Omega}_1$ , and extend the definition to  $\hat{\Omega}_n$  in the obvious manner.

### 3. Group waiting time.

We shall study the group waiting time process defined as follows:

#### Definition 3.1

By the group waiting time  $W_n$ ,  $n = 0, 1, 2, \dots$  we mean the waiting time (excluding service time) of the first customer from  $C_n$  who is taken into service.

It is convenient to work with the random variable  $Z_n$  which is defined as the time  $C_n$  spends in the queue until there is no one ahead, except the ones being served. If  $L_n$  is the time interval from the instant when  $Z_n$  becomes zero and until the last service group with customers from  $C_n$  starts, then the sequence  $\{Z_n\}$ ,  $n = 0, 1, 2, \dots$  satisfies

$$(3.1) \quad Z_{n+1} = (Z_n + L_n - A_n)^+,$$

with the initial condition  $Z_0 = z$ . We recognize the expression above as an equation of the similar type as the waiting time equation found by Lindley [6]. However, there is an important difference: The random variables  $Z_n$  and  $L_n$  are no longer independent as in the model GI/G/1.

Consider now the servicing of the  $n$ -th arrival group  $C_n$ . Service may be performed in one or several groups. The last service group which has customers from  $C_n$  will be called the  $n$ -th rest batch. The rest batch may be filled with customers from  $C_n$  or there may be places available for customers from  $C_{n+1}$ .

We define two random variables  $T_n$  and  $J_n$ , as follows: If the  $n$ -th restbatch can accept  $s$  customers and contains only  $t \leq s$  customers from  $C_n$ , then  $|J_n| = s-t$ . If the  $n$ -th restbatch contains customers from  $C_{n-1}$ ;  $J_n = -|J_n|$ . Otherwise  $J_n = |J_n|$ .  $T_n$  is the number of customers which can be accepted from  $C_{n+1}$  in the first service group with customers from  $C_{n+1}$ . We define  $J_{-1} = T_{-1}$ ; the capacity of the initial service group.

If  $S(T_{n-1})$  denotes service time of  $C_n$ , excluding the service time of the  $n$ -th rest batch, we realize that  $Z_n > 0$  implies that

$$(3.2) \quad L_n = \begin{cases} S(T_{n-1}) + B\delta(0, T_{n-1}) = S'(T_{n-1}) & \text{when } J_n \geq 0 \\ 0 & \text{when } J_n < 0, \end{cases}$$

and  $T_{n-1} = J_{n-1}$ ,

since  $C_n$  will have to wait an extra service period when  $J_{n-1} = 0$ ,  $J_n \geq 0$ . When  $Z_n = 0$  the first service group from  $C_n$  has capacity  $Y$  and  $C_n$  must wait for the time  $W_n$  before service starts, hence

$$(3.3) \quad L_n = \begin{cases} S(T_{n-1}) + W_n & \text{when } J_n \geq 0 \\ 0 & \text{when } J_n < 0, \end{cases}$$

and  $T_{n-1} = Y$ .

We define matrices of distribution functions

$$U_n(t) = \{U_n^i(t)\delta(j,0)\}, \quad V_n(t) = \{V_n^i(t)\delta(j,0)\}, \quad K(t) = \{K_{ij}\}$$

and  $H(t) = \{H_{ij}\}$

for  $i, j = -(m-2), -(m-1), \dots, m-1$ ,

by

$$U_n^i(t) = \Pr\{Z_n \leq t, J_{n-1} = i\},$$

$$V_n^i(t) = \Pr\{W_n \leq t, J_{n-1} = i\},$$

$$K_{ij} = \begin{cases} \Pr\{S'(T_{n-1}) \leq t, J_n=i | J_{n-1}=j, Z_n > 0\} & \text{when } i \geq 0 \\ \Pr\{0 \leq t, J_n=i | J_{n-1}=j\} & \text{when } i < 0, \end{cases}$$

$$H_{ij} = \begin{cases} \Pr\{S'(T_{n-1}) \leq t, J_n=i | J_{n-1}=j, Z_n=0\} & \text{when } i \geq 0 \\ 0 & \text{when } i < 0. \end{cases}$$

If  $\mu$  is the probability distribution of a random variable  $X$ ,  $\mu^*$  denote the distribution of  $(-X)$ .

Let

$$M_n = Z_n + S'(J_{n-1}) - A_n, N_n = W_n + S(Y) - A_n, D_{ij} = \{J_n=i, J_{n-1}=j\}.$$

Considering the possible events between the  $n$ -th and the  $n+1$ -th arrival we find

$$(3.8 \text{ a}) \quad U_{n+1}^i(t) = \sum_j \Pr\{(Z_n - A_n)^+ \leq t, Z_n > 0, D_{ij}\} \quad \text{when } i < 0,$$

$$(3.8 \text{ b}) \quad U_{n+1}^i(t) = \sum_j [\Pr\{M_n^+ \leq t, Z_n > 0, D_{ij}\} + \Pr\{N_n^+ \leq t, Z_n = 0, D_{ij}\}]$$

when  $i \geq 0$ ,

$$(3.9 \text{ a}) \quad V_{n+1}^i(t) = \sum_j [\Pr\{(Z_n + \delta(0, J_n)B - A_n)^+ \leq t, Z_n - A_n > 0, D_{ij}\}]$$

$$+ \Pr\{(Z_n + B - A_n)^+ \leq t, Z_n - A_n \leq 0, Z_n > 0, D_{ij}\}]$$

when  $i < 0$ . (3.9 a) can be written



$$\begin{aligned}
 V_{n+1}^i(t) &= \sum_j [\Pr\{(Z_n - A_n)^+ + \delta(O, J_n)B \leq t, D_{ij}\} \\
 &- \Pr\{\delta(O, J_n)B \leq t, (Z_n - A_n)^+ = 0, D_{ij}\} \\
 &+ \Pr\{(Z_n - A)^+ + B \leq t, Z_n > 0, D_{ij}\} \\
 &- \Pr\{B \leq t, (Z_n - A_n)^+ = 0, Z_n > 0, D_{ij}\}] \\
 &= ((b^{\delta(O, i)} - b)(U_{n+1}^i - \epsilon U_{n+1}^i(0)))(t) \\
 &+ \sum_j \Pr\{(Z_n - A_n + B)^+ \leq t, Z_n > 0, D_{ij}\}.
 \end{aligned}$$

When  $i \geq 0$  the expression is more complicated;

$$\begin{aligned}
 (3.9 \text{ b}) \quad V_{n+1}^i(t) &= \sum_j [\Pr\{M_n + B\delta(O, J_n) \leq t, M_n > 0, Z_n > 0, D_{ij}\} \\
 &+ \Pr\{(M_n + B)^+ \leq t, M_n \leq 0, Z_n > 0, D_{ij}\} \\
 &+ \Pr\{N_n + B\delta(O, J_n) \leq t, N_n > 0, Z_n = 0, D_{ij}\} \\
 &+ \Pr\{(N_n + B)^+ \leq t, N_n \leq 0, Z_n = 0, D_{ij}\}].
 \end{aligned}$$

If (3.9 b) is rewritten in the same way as (3.9 a) we obtain

$$\begin{aligned}
 V_{n+1}^i(t) &= ((b^{\delta(O, i)} - b)(U_{n+1}^i - \epsilon U_{n+1}^i(0)))(t) \\
 &+ \sum_j [\Pr\{(M_n + B)^+ \leq t, Z_n > 0, D_{ij}\} \\
 &+ \Pr\{(N_n + B)^+ \leq t, Z_n = 0, D_{ij}\}], \quad i \geq 0.
 \end{aligned}$$

Since  $Z_n > 0$  implies  $W_n = Z_n + B\delta(O, J_{n-1})$ , the probability of  $\{(N_n + B)^+ \leq t, Z_n = 0\}$  can be written

$$\Pr\{(N_n + B)^+ \leq t, Z_n = 0\} = \Pr\{(N_n + B)^+ \leq t\}$$

$$- \Pr\{(Z_n + S(Y) + (1 + \delta(O, J_{n-1}))B - A_n)^+ \leq t, Z_n > 0\},$$

whence

$$(3.10 \text{ a}) \quad U_{n+1}^i = \sum_j T(a * K_{ij} (U_n^j - \epsilon U_n^j(O))) \quad \text{when } i < 0,$$

$$(3.10 \text{ b}) \quad U_{n+1}^i = \sum_j [T(a * (K_{ij} - b^{\delta(O, j)} H_{ij})) (U_n^j - \epsilon U_n^j(O))] \\ + T(a * H_{ij} V_n^j)] \quad \text{when } i \geq 0$$

and

$$(3.10 \text{ c}) \quad V_{n+1}^i = (b^{\delta(O, i)} - b) (U_{n+1}^i - \epsilon U_{n+1}^i(O)) \\ + \sum_j T(a * b K_{ij} (U_n^j - \epsilon U_n^j(O))) \quad \text{when } i < 0,$$

$$(3.10 \text{ d}) \quad V_{n+1}^i = (b^{\delta(O, i)} - b) (U_{n+1}^i - \epsilon U_{n+1}^i(O)) \\ + \sum_j [T(a * b (K_{ij} - b^{\delta(O, j)} H_{ij})) (U_n^j - \epsilon U_n^j(O))] \\ + T(a * b H_{ij} V_n^j)] \quad \text{when } i \geq 0.$$

Hence we have established a set of equation for  $U_n$  and  $V_n$ .

Lemma 3.1

$$(i) \quad K_{ij} = \begin{cases} f(|j|+i) , & \text{when } i < 0 \\ \sum_{r \geq 1} b^r \sum_{p > 0} (f(g^*)^{r-1})(p+|j|)g(i+p) + \delta(0,i)f(|j|)e & \end{cases}$$

when  $i \geq 0$  ,

$$(ii) \quad H_{ij} = \begin{cases} \sum_{r \geq 0} b^r \sum_{p > 0} (f(g^*)^r)(p)g(i+p) , & \text{when } i \geq 0 \\ 0 & \text{when } i < 0 , \end{cases}$$

(iii)  $bH_{ij} = K_{io}$  when  $i \geq 0$  .

Proof:

Let  $R(J_{n-1})+1$  denote the number of service groups with customers from  $C_n$  . When  $J_n \geq 0$  ,  $J_{n-1} = j \neq 0$  ,  $Z_n > 0$  ,  $R(J_{n-1})$  must satisfy

$$(3.11 a) \quad J_n + X = \sum_{s=1}^{R(j)} Y_s + |j| , \quad X > \sum_{s=1}^{R(j)-1} Y_s + |j|$$

because

$$\sum_{s=1}^{R(j)} Y_s + |j|$$

customers are served in  $R(j)+1$  groups and the rest of  $C_n$  is served in the  $n$ -th restbatch. When  $J_{n-1} = 0$  , the restbatch is complete and  $T_{n-1} = Y$  , so that

$$(3.11 b) \quad J_n + X = \sum_{s=1}^{R(0)} Y_s + Y , \quad X > \sum_{s=1}^{R(0)-1} Y_s + Y .$$

By an elementary argument

$$\Pr\{R(J_{n-1})=r, J_n=i | J_{n-1}=j\} = \sum_{p>0} (f(g^*)^{r-1})(p+|j|)g(i+p),$$

$$i \geq 0, j \neq 0, r \geq 1,$$

$$\Pr\{R(J_{n-1})=r, J_n=i | J_{n-1}=0\} = \sum_{p>0} (f(g^*)^r)(p)g(i+p), \quad i \geq 0, r \geq 0,$$

$$\Pr\{S'(J_{n-1}) \leq t | R(J_{n-1})=r, J_n=i | J_{n-1}=j, Z_n > 0\} = b^{r+\delta(0,j)}(t).$$

When  $J_n < 0, Z_n > 0$ , the relation

$$X + |J_n| = |J_{n-1}|$$

must be valid. Furthermore  $S(Y)$  is seen to have the same distribution as  $S(0)$ . The theorem now follows easily.

It is convenient to introduce some further matrix notations.

Let  $F = \{F_{ij}\}$ ,  $G = \{G_{ij}\}$ ,  $E = \{E_{ij}\}$ ,  $K' = \{K'_{ij}\}$  and  $v_r = \{v_r^{ij}\}$  be the matrices with entries

$$F_{ij} = f(j-i), \quad K'_{ij} = \sum_{r \geq 0} b^r \sum_{p > 0} (f(g^*)^r)(p+j)g(i+p),$$

for  $i, j = 0, 1, 2, \dots, m-1$ ,

and  $F_{ij} = K'_{ij} = 0$  for  $i, j = m, \dots, m+(l-r)^+-1$ .

$$G_{ij} = g(i-j), \quad E_{ij} = \sum_{p > 0} g(i+p)f(p+j), \quad \text{for } i, j = 0, 1, \dots, m+(l-m)^+-1,$$

$$v_r^{ij} = \delta(i, 0) \quad \text{for } i, j = 0, 1, \dots, r-1.$$

Lemma 3.2

$$(i) \quad K' = E(\epsilon I_m - bG)^{-1} I_m$$

$$(ii) \quad K_{ij} = \begin{cases} \epsilon F_{-ij} , & \text{when } i < 0, j \geq 0 \\ bK_{ij} + \delta(0,i) F_{ij} e , & \text{when } i \geq 0, j \geq 0 . \end{cases}$$

Proof:

(ii) is seen immediately.

(i): Since  $EG^r = \{(EG^r)_{ij}\} = \{\sum_{p>0} (f(g^*)^r)(p+j)g(i+p)\}$ ,

$i, j = 0, 1, \dots, m-1$ , the theorem follows. Observe that  $G^r = 0$  when  $r > m+(1-m)^+$ , hence there is no convergence problem.

Remark:

Even if the matrices involved are defined  $m$  dimensional they are understood to be  $m+(n-m)^+$  dimensional with the undefined entries equal to zero.

By the substitutions  $Q_n = \{Q_n^{ik}\}$ ,  $P_n = \{P_n^{ik}\}$ , where

$$(3.12 \text{ a}) \quad Q_n^{ik} = (V_n^i + (b-b^{\delta(0,i)})(U_n^i - \epsilon U_n^i(0)))\delta(k,0)$$

and

$$(3.12 \text{ b}) \quad P_n^{ik} = (U_n^{-i} + U_n^i - \delta(0,i) \sum_r (U_n^{-r} + U_n^r))\delta(k,0) ,$$

$$i, k = 0, 1, \dots, m-1 ,$$

it is possible to write (3.10) on the form

$$(3.13 \text{ a}) \quad U_{n+1} = T(a^* \begin{bmatrix} 0 \\ K' \end{bmatrix} v_{2m-1} Q_n) + T(a^* \begin{bmatrix} F \\ bK' \end{bmatrix} (P_n - \epsilon P_n(0))),$$

$$(3.13 \text{ b}) \quad Q_{n+1} = T(a^* b \begin{bmatrix} 0 \\ K' \end{bmatrix} v_{2m-1} Q_n) + T(a^* b \begin{bmatrix} F \\ bK' \end{bmatrix} (P_n - \epsilon P_n(0))),$$

whence theorem 3.3 follows.

Theorem 3.3

Let

$$\tilde{P}_n = \begin{bmatrix} P_n \\ v_{2m-1} Q_n \end{bmatrix}$$

and assume that  $U_0(t) = I_{2m-1} \epsilon(z+t)$ ,  $V_0(t) = I_{2m-1} \epsilon(w+t)$ .

Then  $U_n$  and  $V_n$  are uniquely determined by (3.13) and

$$(3.14) \quad \tilde{P}_{n+1} = T(a^* K_1 (\tilde{P}_n - \epsilon \tilde{P}_n(0))) + T(a^* K_2) \tilde{P}_n(0), \quad n = 0, 1, \dots,$$

where

$$K_1 = \begin{bmatrix} (I_m - v_m)(F + bK'), (I_m - v_m)K' \\ bv_m(F + bK'), bv_m K' \end{bmatrix}, \quad K_2 = \begin{bmatrix} 0, (I_m - v_m)K' \\ 0, bv_m K' \end{bmatrix}.$$

Equation (3.14) is called the waiting time equation.

Corollary 3.4

When the traffic intensity  $\rho = E(X)E(B)/E(Y)E(A)$  is less than unity the stationary distribution  $\tilde{P} = \lim_{n \rightarrow \infty} \tilde{P}_n$  exists and is determined by

$$(3.15) \quad \tilde{P} = T(a^* K_1 (\tilde{P} - \epsilon \tilde{P}(0))) + T(a^* K_2) \tilde{P}(0).$$

A proof of the existence of the stationary distribution is given in [7] and [2].

We will assume that the customers in an arrival group are ordered and the queuedisiplin is first come first served.

Let  $W_n^p$  be the waiting time of customer nr.  $p$  in  $C_n$  given that  $C_n$  contains at least  $p$  customers. Let

$$\Gamma_n^p = \Pr\{W_n^p \leq t\} .$$

The problem of finding  $\Gamma_n^p$  can be solved by the following argument: If  $C_n$  contains exactly  $p$  customer then obviously

$$\Gamma_n^p = \sum_i U_{n+1}^i .$$

Hence we must have

$$\Gamma_n^p = T(a * v_m K'_p v_{2m-1} Q_n) + T(a * v_{2m-1} K_p (P_n - e P_n(0))) ,$$

where  $K'_p$  and  $K_p$  are  $K'$  and  $K$  respectively, when  $f(\cdot)$  is replaced by  $\delta(p, \cdot)$  .

#### 4. The model $GI^X/E_k^Y/1$ .

Within this model it is possible to obtain solutions of the waiting time equation (by using the approach suggested in [3] p.p. 312-313). Let  $e_\theta$ ;  $\theta \in \mathbb{C}$  , denote the complex exponential measure defined by

$$e_\theta(E) = \int_{E \cap \mathbb{R}^+} \theta \exp(-\theta x) dx , \quad \text{if } \operatorname{Re} \theta > 0 ,$$

$$e_\theta(E) = \int_{E \cap -\mathbb{R}^+} \theta \exp(-\theta x) dx , \quad \text{if } \operatorname{Re} \theta < 0 ,$$

The following lemma is proved in the same way as the analogous results in [3] p.p. 312-313.

Lemma 4.1

$$(i) \quad e_{\theta} e_{\mu}^n = \beta^n e_{\theta} + (1-\beta) \sum_{k=1}^n \beta^{n-k} e_{\mu}^k,$$

where

$$\beta = \frac{\mu}{\mu-\theta}.$$

$$(ii) \quad T(a * e_{\mu}^n) = \sum_{k=1}^n (e_{\mu}^k - \epsilon) \gamma_{n-k} + \epsilon,$$

where

$$\gamma_m = \int_0^{\infty} \frac{e^{-\mu y} (\mu y)^m}{m!} a(dy).$$

By the introduction of the generating functions

$$\psi_X(t) = \sum_{n \geq 0} P_n(t) x^n, \quad \varphi_X(t) = \sum_{n \geq 0} v_{2n-1} Q_n(t) x^n.$$

$$\tilde{\psi}_X = \begin{pmatrix} \psi_X \\ \varphi_X \end{pmatrix} = \sum_{n \geq 0} \tilde{P}_n x^n.$$

(3.14) is transformed to the equivalent equation

$$(4.1) \quad \tilde{\psi}_X = xT(a * K_1(\tilde{\psi}_X - \epsilon \tilde{\psi}_X(0))) + xT(a * K_2) \tilde{\psi}_X(0) + \tilde{P}_0.$$



By the transformation

$$\omega_x = (\epsilon I_m - bG)^{-1} (b(\psi_x - \psi_x(0)\epsilon) + \varphi_x)$$

we get

$$(4.2 \text{ i}) \quad \psi_x = x(I_m - \nu_m)T(a*(F(\psi_x - \epsilon\psi_x(0)) + E\omega_x)) + P_0,$$

$$(4.2 \text{ ii}) \quad \omega_x - b(G\omega_x + \psi_x - \epsilon\psi_x(0)) = x\nu_m T(a*b(F(\psi_x - \epsilon\psi_x(0)) + E\omega_x) + \nu_m Q_0).$$

By assumption  $b = e_{\mu}^k$ . Assume that (4.2) has a solution of hyperexponential type; i.e.,

$$(4.3) \quad \begin{pmatrix} \psi_x \\ \omega_x \end{pmatrix} = \begin{pmatrix} q_0' \\ q_0'' \end{pmatrix} \epsilon + \sum_{j \geq 1}^h \begin{pmatrix} q_j' \\ q_j'' \end{pmatrix} e_{\theta_j}.$$

If  $U_0 = \epsilon U_0(0)$  and  $V_0 = \epsilon V_0(0)$ , it follows from the definition of  $\tilde{P}_n$  that  $\tilde{P}_0 = \begin{pmatrix} (I_m - \nu_m)U_0 \\ \nu_m V_0 \end{pmatrix}$ .

Inserting (4.3) in (4.2) gives

$$\epsilon q_0' + \sum_{j > 0} e_{\theta_j} q_j' = x(I_m - \nu_m) \sum_{j > 0} T(a* e_{\theta_j} (Fq_j' + Eq_j'')) + x(I_m - \nu_m) Eq_0'' \epsilon + (I_m - \nu_m) U_0,$$

$$(\epsilon I_m - bG)q_0'' + \sum_{j > 0} (e_{\theta_j} q_j'' - e_{\theta_j} b(Gq_j'' + q_j'))$$

$$= x\nu_m \sum_{j > 0} T(a*b e_{\theta_j} (Fq_j' + Eq_j'')) + x\nu_m T(a*b Eq_0'') + \nu_m V_0.$$

- From Lemma 4.1 follows

$$(4.4 \text{ i}) \quad \epsilon q'_0 + \sum_{j>0} e_{\theta_j} q'_j - (I_m - \nu_m) x \sum_{j>0} (Fq'_j + Eq''_j) (\alpha_j e_{\theta_j} + (1 - \alpha_j) \epsilon) \\ - x (I_m - \nu_m) Eq''_0 \epsilon + (I_m - \nu_m) U_0 = 0 ,$$

$$(4.4 \text{ ii}) \quad \sum_{j>0} ((\epsilon I_m - \beta_j^k G) q''_j - \beta_j^k q'_j) e_{\theta_j} - (e_{\mu}^k - \epsilon) Gq''_0 - Gq''_0 \epsilon + q''_0 \epsilon \\ - \sum_{j>0} (Gq''_j + q'_j) ((1 - \beta_j) \sum_{r=1}^k \beta_j^{k-r} (e_{\mu}^r - \epsilon) + (1 - \beta_j^k) \epsilon) \\ - x \nu_m \sum_{j>0} (Fq'_j + Eq''_j) \beta_j^k (\alpha_j e_{\theta_j} + (1 - \alpha_j) \epsilon) \\ - x \nu_m \sum_{j>0} (Fq'_j + Eq''_j) ((1 - \beta_j) \sum_{r=1}^k \beta_j^k \eta_{rj} (e_{\mu}^r - \epsilon) + (1 - \beta_j^k) \epsilon) \\ - x \nu_m Eq''_0 \left( \sum_{r=1}^k (e_{\mu}^r - \epsilon) \gamma_{k-r} + \epsilon \right) - \nu_m V_0 = 0 ,$$

where

$$\eta_{rj} = \sum_{s \geq r} \beta_j^{-r} \gamma_{s-r} , \quad \alpha_j = \alpha(\theta_j) .$$

The equations above imply that the coefficients of  $\epsilon$ ,  $(e_{\mu}^r - \epsilon)$ ,  $r = 1, 2, \dots, k$ , and  $e_{\theta_j}$ ,  $j = 1, 2, \dots$  are zero. Hence, we are led to the equations

$$(4.5 \text{ i}) \quad q'_0 = \sum_{j>0} \frac{1 - \alpha_j}{\alpha_j} q'_j + x (I_m - \nu_m) Eq''_0 + (I_m - \nu_m) U_0 ,$$

$$(4.5 \text{ ii}) \quad q'_j = \beta_j^{-k} (I_m - \nu_m) (I_m - \beta_j^k G) q''_j ,$$

$$(4.5iii) \quad (I_m - G - x v_m E) q_0'' = \sum_{j>0} \left( \frac{1 - \alpha_j}{\alpha_j \beta_j^k} v_m (I - \beta_j^k G)^{-1 + \beta_j^{-k}} \right) q_j'' + v_m V_0 ,$$

$$(4.5iv) \quad (\delta(r, k) G + x \gamma_{k-r} v_m E) q_0'' = \sum_{j>0} [(\beta_j - 1) \beta_j^{-r} ,$$

$$+ (\beta_j - 1) (\eta_{rj} \alpha_j^{-1} - \beta_j^{-r}) v_m (I_m - \beta_j^k G)] q_j'' , \quad r = 1, 2, \dots, k ,$$

$$(4.5 v) \quad ((I_h - x \alpha_j F) (I_m - \beta_j^k G) - x \alpha_j \beta_j^k E) q_j'' = 0 , \quad j = 1, 2, \dots ,$$

where  $\beta_j, j = 1, 2, \dots, hk$  , are the roots of

$$(4.5vi) \quad \det\{(I_h - x \alpha F) (I_m - \beta^k G) - x \alpha \beta^k E\} ,$$

where  $h = m + (l - m)^+$ .

Observe that (4.5 iv-v) constitutes a set of  $2kh$  equations. Hence it is sufficient that (4.5 vi) has  $kh$  roots.

### 5. Special cases.

Unfortunately, the assumptions (4.3) are not always fulfilled. For instance when  $Y = m > 1$  . However, in this case we are able to slighen the assumptions (4.3). Observe that  $Y = m > 1$  implies  $G = 0$  . Let

$$(5.1) \quad \xi_x = (F + bE) (\psi_x - \epsilon \psi_x(0)) + E \rho_x$$

whence

$$(5.2a) \quad \psi_x = x (I_m - v_m) T(a * \xi_x) + (I_m - v_m) U_0 ,$$

$$(5.2b) \quad \varphi_x = x v_m T(a * b \xi_x) + v_m V_0$$

and

$$(5.3) \quad \xi_x = x(F+bE)(I_m - v_m)(Ta * \xi_x - eT(a * \xi_x)(0)) + xEv_m T(a * b \xi_x) + Ev_m V_0 .$$

We now suppose that (5.3) has a solution of the form

$$\xi_x = p_0 e + \sum_{j>0} e_{\theta_j} p_j ,$$

that is;  $p_0, p_j, j > 0$  , must satisfy

$$p_0 e + \sum_{j>0} e_{\theta_j} p_j = x(F+bE)(I_m - v_m) \sum_{j>0} T(a * e_{\theta_j}) p_j + xEv_m \sum_{j>0} T(a * b e_{\theta_j}) p_j + xEv_m T(a * b) p_0 + Ev_m V_0 .$$

Exactly the same calculations as in the preceding section give

$$(5.4 i) \quad p_j = x \alpha_j (F + \beta_j^k E) p_j , \quad j = 1, 2, \dots ,$$

$$(5.4ii) \quad \sum_{j>0} \{ (\beta_j - 1) \beta_j^{k-r} \alpha_j^{-\delta(r,k)} (1 - \alpha_j) \} E(I_m - v_m)$$

$$+ (\beta_j - 1) \beta_j^k \eta_{rj}^{Ev_m} \} p_j = \gamma_{k-r}^{Ev_m} p_0 , \quad r = 1, 2, \dots, k ,$$

$$(5.4iii) \quad (I_m - xEv_m) p_0 = \sum_j \{ (\alpha_j^{-1} - 1) (I_m - v_m) + xEv_m + x \alpha_j E(I_m - v_m)$$

$$- x \alpha_j \beta_j^k E \} p_j + Ev_m V_0 .$$

We shall now consider the case when both the arrival and the service group capacity are of constant size; i.e.  $X = 1$ ,  $Y = m$ . Assume first that  $m < 1$ . Then  $F = 0$  and

$$I_1 - \beta^k G - x\alpha \beta^k E = I_1 - \beta^k \begin{bmatrix} 0 & , x\alpha I_m \\ I_{1-m}, & 0 \end{bmatrix} .$$

It follows that

$$(5.5) \quad \det(I_1 - \beta^k G - x\alpha \beta^k E) = 1 - (x\alpha)^m \beta^{kl} .$$

From Lemma 1 in Takács [8] page 82 we conclude that (5.5) has  $kl$  distinct roots  $\beta_j = \beta(\theta_j)$ ,  $j = 1, 2, \dots, kl$  for  $\text{Re} \theta \geq 0$ . Hence equations (4.5) have a solution. Assume that  $1 \leq m$ . Then  $G = 0$ ;

$$I_m - x\alpha (F + \beta^k E) = I_m - x\alpha \begin{bmatrix} 0 & , I_{m-1} \\ \beta^k I_1, & 0 \end{bmatrix} ,$$

and

$$(5.6) \quad \det(I_m - x\alpha F - x\alpha \beta^k E) = 1 - (x\alpha)^m \beta^{kl} .$$

When  $p_0$  is eliminated, (5.4ii) is a set of  $kl$  equations because

$$E = \begin{bmatrix} 0 & , & 0 \\ I_1 & , & 0 \end{bmatrix} .$$

Accordingly, (5.4) has a solution since (5.6) has  $kl$  distinct roots.

The stationary solution is obtained by multiplying (4.5) and (5.4) by  $1-x$  and let  $x$  tend to  $1^-$ .

6. The model  $E_k^X/G^Y/1$  .

The "key" to the solution is the fact that the assumption  $a = e_\lambda^k$  enables us to express the operator  $T$  in a special form

Lemma 6.1

Suppose  $\mu \in \Omega_m^+$  . Then

$$T((e^*)^k \mu) = (e^*)^k \mu + (\epsilon - e^*) \sum_{r=1}^k \alpha_r (e^*)^{k-r} ,$$

where

$$\alpha_r = ((e^*)^r \mu)(\sim \mathbb{R}^+) .$$

Proof:

For  $k = 1$  the Lemma reduces to equation (72) in Kingman [3]. Assume (6.1) valid for  $k = 1, 2, \dots, p$  . Since  $T(e^*)^r = \epsilon$  we get

$$T((e^*)^{p+1} \mu) = T(e^* T((e^*)^p \mu)) .$$

Now  $\nu = T((e^*)^p \mu) \in \Omega_m^+$  implies

$$T((e^*)^{p+1} \mu) = T(e_\lambda^* \nu) = e^* \nu + (\epsilon - e^*) \alpha' ,$$

where  $\alpha' = (e^* \nu)(\sim \mathbb{R}^+) .$

Obviously  $T(e^* \nu)(\sim \mathbb{R}^+) = (e^* \nu)(\sim \mathbb{R}^+)$  which yields  $\alpha' = \alpha_{p+1} .$

Hence, by the assumption

$$T(e^* \nu) = (e^*)^{p+1} \mu + (\epsilon - e^*) \sum_{r=1}^p \alpha_r (e^*)^{p+1-r} + (\epsilon - e^*) \alpha_{p+1}$$

and the Lemma follows by induction.

With  $\mu = K_1 \tilde{\psi}_x$  and  $e = e_\lambda$  application of (6.1) on (4.1) gives (with  $I = I_{2m-1}$ )

$$(6.2) \quad (\epsilon I - xa * K_1) \tilde{\psi}_x = x(\epsilon - e*) \sum_{r=1}^k \alpha_r (e*)^{k-r} + xT(a*(K_2 - K_1)) \tilde{\psi}_x(0) + \tilde{P}_0 .$$

Let  $\Delta_x = \epsilon I - xa * K_1$  .

Inserting  $t = 0$  gives an expression for  $\alpha_k$  , viz.,

$$(6.3) \quad x\alpha_k = (\epsilon I - x(a*(K_2 - K_1))(0)) \tilde{\psi}_x - \tilde{P}_0(0) .$$

Since  $\|a*K_1\| \leq 1$  ,  $\Delta_x^{-1} = \sum_{r \geq 0} (xa*K_1)^r$  exists when  $|x| < 1$  .

Thus (6.3) and (6.2) are equivalent to

$$(6.4) \quad \tilde{\psi}_x - \epsilon \tilde{\psi}_x(0) = \Delta_x^{-1} (\tilde{P}_0 - (\epsilon - e_\lambda *) \tilde{P}_0(0)) + (\epsilon - e_\lambda *) \Delta_x^{-1} \sum_{r=1}^{k-1} \alpha_r (e_\lambda *)^{k-r} - e_\lambda * \Delta_x^{-1} (I \epsilon - x(e_\lambda *)^{k-1} K_2) \tilde{\psi}_x(0) .$$

If  $k = 1$  ,  $t = 0$  determines  $\tilde{\psi}_x(0)$  by

$$(6.5) \quad (e_\lambda * \Delta_x^{-1} (\epsilon I - x(e_\lambda *)^{k-1} K_2)(0)) \tilde{\psi}_x(0) = (\Delta_x^{-1} (\tilde{P}_0 - (\epsilon - e_\lambda *) \tilde{P}_0(0)))(0) .$$

When  $k > 1$  let

$$c_r = (\epsilon - e_\lambda *) (e_\lambda *)^{k-r} , \quad r = 1, 2, \dots, k-1 ,$$

$$C_x = e_\lambda * (\epsilon I - x(e_\lambda *)^{k-1} K_2) ,$$

$$D = (\tilde{P}_0 - (\epsilon - e_\lambda *) \tilde{P}_0(0)) ,$$

$$E_j = (e_\lambda *)^j K_1 \Delta_x^{-1} .$$

Thus, (6.4) can be written

$$(6.6) \quad \tilde{\Psi}_X = (\epsilon I - \Delta_X^{-1} C_X) \tilde{\Psi}_X(0) + \sum_{r=1}^{k-1} c_r \Delta_X^{-1} \alpha_r + \Delta_X^{-1} D .$$

Furthermore  $(E_j \tilde{\Psi}_X)(0) = \alpha_j$  leads to

$$(6.7 \text{ i}) \quad \alpha_j = (E_j (\Delta_X^{-1} - C_X))(0) \tilde{\Psi}_X(0) + \sum_{r=1}^{k-1} (E_j c_r)(0) \alpha_r + E_j D ,$$

$$j = 1, 2, \dots, k-1 ,$$

$$(6.7 \text{ ii}) \quad (\Delta_X^{-1} C_X)(0) \tilde{\Psi}_X(0) = \sum_{r=1}^{k-1} c_r(0) \alpha_r + (E_j D)(0)$$

which determines  $\alpha_j$ ,  $j = 1, 2, \dots, k-1$ , and  $\tilde{\Psi}_X(0)$  .

## 7. The stationary solution

In this section we shall demonstrate how the stationary solution of (6.2) can be obtained by use of the Fourier-Stieltjes transform. In the stationary case we have

$$(7.1) \quad (\epsilon I - a * K_1) \tilde{P} = (\epsilon - e_\lambda *) \sum_{r=1}^k \alpha_r (e_\lambda *)^{k-r} + T(a * (K_2 - K_1)) \tilde{P}(0) ,$$

where now

$$\alpha_r = (e_\lambda *^r K_1 \tilde{P})(0) .$$

After the introduction of the Fourier transform an equation analogous to (6.6) is obtained

$$(7.2) \quad \hat{\Delta}_1(z) (\hat{P}(z) - \tilde{P}(0)) = -\hat{C}_1(z) \tilde{P}(0) + \sum_{r=1}^{k-1} \hat{c}_r(z) \alpha_r .$$



Let  $\kappa$  be defined by

$$\hat{\kappa}\hat{\Delta}_1 = \det \hat{\Delta}_1,$$

whence

$$(7.3) \quad \det(\Delta_1(z))(\tilde{P}(z)-\tilde{P}(0)) = -\hat{\kappa}(z)\hat{C}_1(z)\tilde{P}(0) + \sum_{r=1}^{k-1} \hat{c}_r(z)\hat{\kappa}(z)\alpha_r.$$

Suppose that  $\det \Delta_1(z)$  has  $k$  roots  $z_1, z_2, \dots, z_k$ . Then  $\tilde{P}(0), \alpha_1, \dots, \alpha_{k-1}$ , are determined except for a constant by

$$(7.4) \quad \hat{\kappa}(z_i)\hat{C}_1(z_i)\tilde{P}(0) = \sum_{r=1}^{k-1} \hat{\kappa}(z_i)\hat{c}_r(z_i)\alpha_r, \quad i = 1, 2, \dots, k.$$

Premultiplicating (7.2) by

$$v = \begin{bmatrix} v_m, 0 \\ 0, v_m \end{bmatrix}$$

gives an equation where both sides become zero when  $z = 0$ .

By l'Hospitals rule,

$$(7.5) \quad (\hat{a} * v \hat{K}_1)'(0)(\tilde{P}-\tilde{P}(0)) = v \hat{C}_1'(0)\tilde{P}(0) - i \sum_{r=1}^{k-1} v \alpha_r.$$

Consider  $(\hat{a} * v \hat{K}_1)'(0)\tilde{P}(0)$ .

The process  $\{J_n\}$  is recurrent and therefore  $J_n$  converges in distribution to  $J$ , say. Since

$$v K_1 = (b v_m K, v_m (K-F))$$

and

$$v_m K(t) = (k_0(t), k_1(t), \dots, k_{m-1}(t)) ,$$

where

$$k_j(t) = \sum_i k_{ij}(t) ,$$

it follows that

$$(\hat{a} * \hat{b} v_m \hat{K})'(0) = i(E(S'(j)) | J=j) - iE(A) + iE(B) .$$

Thus

$$\begin{aligned} (\hat{a} * v_m \hat{K}_1)'(0) \hat{P}(0) &= v_m (\hat{a} * \hat{b} \hat{K})'(0) \hat{P}(0) + v_m (\hat{a} * \hat{K})'(0) v_{2m-1} \hat{Q}(0) \\ &= i \sum_j E(S'(j)) | J=j \Pr(J=j) - iE(A) + iE(B) . \end{aligned}$$

By (3.11) we find

$$E(S'(J) - A + B) = (\rho - 1)E(A) + E(B) = k\lambda^{-1}(\rho - 1) + E(B) .$$

Equation (6.12) therefore reduces to

$$(7.6) \quad E(B) - k(1 - \rho) = -i\lambda (\hat{a} * \hat{b} \hat{K})'(0) P(0) + v_{2m-1} Q(0) - \lambda \sum_{r=1}^{k-1} v \alpha_r .$$

Together with (7.4) we have a set of  $k+1$  matrix equations to determine the  $k+1$  unknowns  $\tilde{P}(0), \alpha_1, \dots, \alpha_{k-1}$  .

It is known that a probability distribution can be approximated by a linear combination of Erlang distributions. From Lemma 6.1 it is clear that the results in the last section can be generalized to the case when  $a(\cdot)$  is a linear combination of Erlang distributions. Accordingly, it is possible to obtain approximate solutions of the waiting time equation for general  $a(\cdot)$  .

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