

INVESTIGATION OF G-NETWORK WITH SIGNALS AT TRANSIENT BEHAVIOR

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Abstract. G-queueing network with positive messages and signals at transient behavior is considered. A system of difference-differential equations for the state probabilities of the network is obtained. To find them and the average characteristics of the network a technique was applied based on the use of apparatus of multivariate generating functions. An expression for the generating function was obtained. An example is calculated.

Keywords: *G-queueing network, transient behavior, transient behavior, state probabilities, generating function*

1. General information

Previously were investigated queueing networks (QN) with negative customers (G-network). For such networks an analysis was performed at transient behavior [1, 2]. A system of difference-differential equations (DDE) for the state probabilities was obtained and a solution was found.

The impact of the environment on the queue process of positive messages can benefit not only the negative messages that are simply destroying one or more positive messages in the queueing system QN (QS), but also coming from outside signals-triggers, the effect of which is to move the instant the positive message of this system in some other network system. Thus, the trigger, unlike the negative message, does not destroy the positive message but merely immediately moves it with a specified probability of the system in some other network system. G-network with triggers at the stationary behaviour has been studied in [3, 4]. Note that in some studies [5] the concept of a trigger and signal are identified. Clearly, this is a mixture of terms, however, this fact must be taken into account when studying the works on G-networks.

QS with signals (negative messages and/or triggers) are used in the analytical modeling of information technology systems and networks, with negative messages possibly arising, for example, for simulation of computer viruses, and the introduction of triggers to manage the load on the network. G-network is also widely used to model neural networks, and drive signals are modeled by positive messages,

and braking signals - by negative messages. Further details on the practical use of G-networks with signals are described in [6].

So, consider an open queueing G-network with n single-line QS. In QS S_i from the outside (from the system S_0) an incoming flow of positive (normal) of messages intensity of λ_{0i}^+ and Poisson flow of signals intensity of λ_{0i}^- , $i = \overline{1, n}$. All flows of messages entering the network are independent. The service time of the positive messages in the QS S_i exponentially distributed with mean μ_i , $i = \overline{1, n}$. After servicing the positive message in QS S_i , it is sent to the QS S_j with probability p_{ij}^+ again as positive message, with probability p_{ij}^- - as a signal and with probability $p_{i0} = 1 - \sum_{j=1}^n (p_{ij}^+ + p_{ij}^-)$ is leaving from the network to the external environment, $i, j = \overline{1, n}$.

Thus, in the network are circulating not only positive messages, and signals too. Signal, coming in an empty system S_i (in which there are no positive messages), does not have any impact on the network and immediately disappears from it. Otherwise, if the system S_i is not empty, when it receives a signal, there can occur following events: incoming signal instantly moves the positive message from the system S_i into the system S_j with probability q_{ij} , in this case, signal is referred to as a trigger; or with probability $q_{i0} = 1 - \sum_{j=1}^n q_{ij}$ signal is triggered by a negative message and destroys in QS S_i positive message [3]. The state of the network meaning the vector $k(t) = (k, t) = (k_1, k_2, \dots, k_n, t)$, where k_i - the number of messages at the moment of time t at the system S_i , $i = \overline{1, n}$.

It is necessary to find network state probabilities and average characteristics of the network at transient behavior.

2. Network state probabilities

Lemma. *State probabilities of considered network satisfy the system of DDE:*

$$\begin{aligned} \frac{dP(k, t)}{dt} = & - \sum_{i=1}^n [\lambda_{0i}^+ + (\lambda_{0i}^- + \mu_i)u(k_i)]P(k, t) + \sum_{i=1}^n \{\lambda_{0i}^+ u(k_i)P(k - I_i, t) + \\ & + (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})P(k + I_i, t) + \sum_{j=1}^n [\mu_i (1 - u(k_j))p_{ij}^- P(k + I_i, t) + \\ & + (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})u(k_j)P(k + I_i - I_j, t) + \mu_i p_{ij}^- q_{j0} P(k + I_i + I_j, t) + \\ & + \sum_{s=1}^n \mu_i p_{ij}^- q_{js} u(k_s)P(k + I_i + I_j - I_s, t)\} \end{aligned} \quad (1)$$

where I_i - a vector of dimension n , consisting of zeros, except for the component with number of i , which is equal to 1, $i = \overline{1, n}$; $u(x) = \begin{cases} 1, x > 0 \\ 0, x \leq 0 \end{cases}$ - Heaviside function.

Proof. In view of the exponential service times of messages, a random process $k(t) = (k, t)$ is a Markov chain with a countable number of states. The possible transitions in the state $(k, t + \Delta t)$ for the time Δt :

✓ from the state $(k - I_i, t)$, in this case into QS S_i during time Δt coming positive message with probability $\lambda_{0i}^+ u(k_i) \Delta t + o(\Delta t)$;

✓ from the state $(k + I_i, t)$, while a positive message goes out from the network to the external environment or goes to QS S_j as signal, if it was not messages, or arriving in the QS S_i signal triggered as a negative message and destroys the QS S_i positive message; the probability of such event is equal to $(\mu_i p_{i0} + \lambda_{0i}^- q_{i0} + \mu_i (1 - u(k_j)) p_{ij}^-) \Delta t + o(\Delta t)$;

✓ from the state $(k + I_i - I_j, t)$, in this case, after the servicing of positive message in the QS S_i it goes to the QS S_j again as a positive message or arriving in the QS S_i signal immediately moves the positive message from the system S_i to the system S_j ; the probability of such event is equal to $(\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) u(k_j) \Delta t + o(\Delta t)$;

✓ from the state $(k + I_i + I_j, t)$, while after the servicing of positive message in the QS S_i it goes to the QS S_j as signal, which is triggered as a negative message, destroyed in S_j positive message; the probability of such event is equal to $\mu_i p_{ij}^- q_{j0} \Delta t + o(\Delta t)$;

✓ from the state $(k + I_i + I_j - I_s, t)$, in this case, after the completion of the service message in the QS S_i , it goes to the QS S_j as signal, which immediately moves the positive message from the system S_j to the system S_s ; the probability of such event is equal to $\mu_i p_{ij}^- q_{js} u(k_s) \Delta t + o(\Delta t)$;

✓ from the state (k, t) , while in each QS S_i , $i = \overline{1, n}$, no positive messages arrive, any signals and which during Δt do not to service any message; the probability of such event is equal to $\left(1 - \sum_{i=1}^n [\lambda_{0i}^+ + (\lambda_{0i}^- + \mu_i) u(k_i)] \right) \Delta t + o(\Delta t)$; of the remaining states with a probability $o(\Delta t)$.

Then, using the formula of total probability and taking the limit $\Delta t \rightarrow 0$, we obtain a system of equations for the state probabilities of the network (1). The lemma is proved.

Suppose that all systems of the network are operating in high load, i.e. $k_i(t) > 0$, $\forall t > 0$, $i = \overline{1, n}$, then the system of DDE (1) takes the form

$$\begin{aligned} \frac{dP(k, t)}{dt} = & - \sum_{i=1}^n [\lambda_{0i}^+ + \lambda_{0i}^- + \mu_i] P(k, t) + \\ & + \sum_{i=1}^n \left\{ \lambda_{0i}^+ P(k - I_i, t) + (\mu_i p_{i0} + \lambda_{0i}^- q_{i0}) P(k + I_i, t) + \right. \\ & + \sum_{j=1}^n [(\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) P(k + I_i - I_j, t) + \mu_i p_{ij}^- q_{j0} P(k + I_i + I_j, t) + \\ & \left. + \sum_{s=1}^n \mu_i p_{ij}^- q_{js} P(k + I_i + I_j - I_s, t) \right\}. \end{aligned} \quad (2)$$

Denote by $\Psi_n(z, t)$, where $z = (z_1, z_2, \dots, z_n)$, n -dimensional generating function:

$$\Psi_n(z, t) = \sum_{k_1=0}^{\infty} \sum_{k_2=0}^{\infty} \dots \sum_{k_n=0}^{\infty} P(k_1, k_2, \dots, k_n, t) z_1^{k_1} z_2^{k_2} \dots z_n^{k_n} = \sum_{k_1=0}^{\infty} \sum_{k_2=0}^{\infty} \dots \sum_{k_n=0}^{\infty} P(k, t) \prod_{i=1}^n z_i^{k_i}.$$

Multiplied (2) on $\prod_{l=1}^n z_l^{k_l}$ and adding together all possible values k_l from 1 to $+\infty$, $l = \overline{1, n}$, obtain the inhomogeneous linear differential equations

$$\begin{aligned} \frac{d\Psi_n(z, t)}{dt} = & - \left[\sum_{i=1}^n (\lambda_{0i}^+ + \lambda_{0i}^- + \mu_i) - \sum_{i=1}^n \lambda_{0i}^+ z_i - \sum_{i=1}^n \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} - \right. \\ & - \sum_{i, j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} - \sum_{i, j=1}^n \mu_i p_{ij}^- q_{j0} \frac{1}{z_i z_j} - \sum_{i, j=1}^n \mu_i \sum_{s=1}^n p_{ij}^- q_{js} \frac{z_s}{z_i z_j} \left. \right] \Psi_n(z, t) - \\ & - \sum_{i=1}^n \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} \sum_{\substack{k_j=0 \\ j=1, n, j \neq i}}^{\infty} P(k_1, \dots, k_{i-1}, 0, k_{i+1}, \dots, k_n, t) \prod_{\substack{l=1 \\ l \neq i}}^n z_l^{k_l} - \\ & - \sum_{i, j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} \sum_{\substack{k_m=0 \\ m=1, n, m \neq j}}^{\infty} P(k_1, \dots, k_{i-1}, 0, k_{i+1}, \dots, k_n, t) \prod_{\substack{l=1 \\ l \neq i}}^n z_l^{k_l} - \\ & - \sum_{i, j=1}^n \frac{\mu_i p_{ij}^- q_{j0}}{z_i z_j} \sum_{\substack{k_m=0 \\ m=1, n, m \neq j}}^{\infty} P(k_1, \dots, k_{i-1}, 0, k_{i+1}, \dots, k_n, t) \prod_{\substack{l=1 \\ l \neq i}}^n z_l^{k_l} - \end{aligned} \quad (3)$$

$$- \sum_{i,j=1}^n \mu_i \sum_{s=1}^n p_{ij}^- q_{js} \frac{z_s}{z_i z_j} \sum_{\substack{k_m=0 \\ m=1, n, m \neq j}}^{\infty} P(k_1, \dots, k_{i-1}, 0, k_{i+1}, \dots, k_n, t) \prod_{\substack{l=1 \\ l \neq i}}^n z_l^{k_l}.$$

Since all of the QS network operate under high load conditions, the last four expressions in the form of the sums in equation (3) will be zero, and it becomes homogeneous:

$$\begin{aligned} \frac{d\Psi_n(z, t)}{dt} = & - \left[\sum_{i=1}^n (\lambda_{0i}^+ + \lambda_{0i}^- + \mu_i) - \sum_{i=1}^n \lambda_{0i}^+ z_i - \sum_{i=1}^n \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} - \right. \\ & \left. - \sum_{i,j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} - \sum_{i,j=1}^n \mu_i p_{ij}^- q_{j0} \frac{1}{z_i z_j} - \sum_{i,j=1}^n \mu_i \sum_{s=1}^n p_{ij}^- q_{js} \frac{z_s}{z_i z_j} \right] \Psi_n(z, t). \end{aligned}$$

Its solution has the form

$$\begin{aligned} \Psi_n(z, t) = C_n \exp \left\{ - \left[\sum_{i=1}^n (\lambda_{0i}^+ + \lambda_{0i}^- + \mu_i) - \sum_{i=1}^n \lambda_{0i}^+ z_i - \sum_{i=1}^n \frac{1}{z_i} (\mu_i p_{i0} + \lambda_{0i}^- q_{i0} - \right. \right. \\ \left. \left. - \left(\sum_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) z_j + \mu_i \sum_{j=1}^n \frac{1}{z_j} \left(p_{ij}^- q_{j0} + \sum_{s=1}^n p_{ij}^- q_{js} z_s \right) \right) \right] t \right\}. \end{aligned} \quad (4)$$

We assume that at the initial moment of time network is in state $(\alpha_1, \alpha_2, \dots, \alpha_n, 0)$, $\alpha_i > 0$, $i = \overline{1, n}$,

$$P(\alpha_1, \alpha_2, \dots, \alpha_n, 0) = 1, \quad P(k_1, k_2, \dots, k_n, 0) = 0, \quad \forall \alpha_i \neq k_i, \quad i = \overline{1, n}.$$

Then the initial condition for the last equation (4) will be $\Psi_n(z, 0) = P(\alpha_1, \alpha_2, \dots, \alpha_n, 0) \prod_{l=1}^n z_l^{\alpha_l} = \prod_{l=1}^n z_l^{\alpha_l}$. Using it, we obtain $C_n = 1$.

Thus, the expression for the generating function $\Psi_n(z, t)$ has the form

$$\begin{aligned} \Psi_n(z, t) = a_0(t) \exp \left\{ \sum_{i=1}^n \lambda_{0i}^+ z_i t \right\} \exp \left\{ \sum_{i=1}^n \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} t \right\} \times \\ \times \exp \left\{ \sum_{i,j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} t \right\} \exp \left\{ \sum_{i,j=1}^n \mu_i p_{ij}^- q_{j0} \frac{1}{z_i z_j} t \right\} \times \\ \times \exp \left\{ \sum_{i,j=1}^n \mu_i \sum_{s=1}^n p_{ij}^- q_{js} \frac{z_s}{z_i z_j} t \right\} \prod_{l=1}^n z_l^{\alpha_l} \end{aligned} \quad (5)$$

where $a_0(t) = \exp \left\{ - \sum_{i=1}^n (\lambda_{0i}^+ + \lambda_{0i}^- + \mu_i) t \right\}$.

Transform (5) to a form suitable for finding the state probabilities of the network, expanding its member exhibitors in the Maclaurin series. Then the following statement

Theorem. *The expression for the generating function has the form*

$$\Psi_n(z, t) = a_0(t) \sum_{l_1=0}^{\infty} \dots \sum_{l_n=0}^{\infty} \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} t^{\sum_{i=1}^n (l_i + c_i + r_i + u_i + w_i)} \times$$

$$\times \prod_{i=1}^n \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{c_i} \prod_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})^{r_i} \left(\mu_i \prod_{j=1}^n p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^n \prod_{s=1}^n q_{js} \right)^{w_i}}{l_i! c_i! r_i! u_i! w_i!} \times (6)$$

$$\times z_i^{a_i + l_i - c_i - r_i + R - u_i - W}$$

where $R = \sum_{i=1}^n r_i$, $U = \sum_{i=1}^n u_i$, $W = \sum_{i=1}^n w_i$.

Proof. From relation (5) follows that

$$\Psi_n(z, t) = a_0(t) a_1(z, t) a_2(z, t) a_3(z, t) a_4(z, t) a_5(z, t) \prod_{l=1}^n z_l^{a_l},$$

where:

$$a_1(z, t) = \exp \left\{ \sum_{i=1}^n \lambda_{0i}^+ z_i t \right\} = \prod_{i=1}^n \exp \left\{ \lambda_{0i}^+ z_i t \right\} = \prod_{i=1}^n \sum_{l_i=0}^{\infty} \frac{[\lambda_{0i}^+ z_i t]^{l_i}}{l_i!} =$$

$$= \sum_{l_1=0}^{\infty} \dots \sum_{l_n=0}^{\infty} \prod_{i=1}^n \frac{[\lambda_{0i}^+ z_i t]^{l_i}}{l_i!} = \sum_{l_1=0}^{\infty} \dots \sum_{l_n=0}^{\infty} \frac{t^{l_1 + l_2 + \dots + l_n}}{l_1! l_2! \dots l_n!} \lambda_{01}^+ l_1 \dots \lambda_{0n}^+ l_n z_1^{l_1} \dots z_n^{l_n},$$

$$a_2(z, t) = \exp \left\{ \sum_{i=1}^n \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} t \right\} = \prod_{i=1}^n \exp \left\{ \frac{\mu_i p_{i0} + \lambda_{0i}^- q_{i0}}{z_i} t \right\} =$$

$$= \prod_{i=1}^n \sum_{c_i=0}^{\infty} \frac{[(\mu_i p_{i0} + \lambda_{0i}^- q_{i0}) t z_i^{-1}]^{c_i}}{c_i!} = \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} \prod_{i=1}^n \frac{[(\mu_i p_{i0} + \lambda_{0i}^- q_{i0}) t z_i^{-1}]^{c_i}}{c_i!} =$$

$$= \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} t^{c_1 + \dots + c_n} \frac{(\mu_1 p_{10} + \lambda_{01}^- q_{10})^{c_1} \dots (\mu_n p_{n0} + \lambda_{0n}^- q_{n0})^{c_n}}{c_1! \dots c_n!} z_1^{-c_1} \dots z_n^{-c_n},$$

$$a_3(z, t) = \exp \left\{ \sum_{i,j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} t \right\} = \prod_{i=1}^n \prod_{j=1}^n \exp \left\{ (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) \frac{z_j}{z_i} t \right\} =$$

$$= \prod_{i=1}^n \prod_{j=1}^n \sum_{r_i=0}^{\infty} \frac{[(\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) t z_j z_i^{-1}]^{r_i}}{r_i!} = \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \prod_{i=1}^n \prod_{j=1}^n \frac{[(\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij}) t z_j z_i^{-1}]^{r_i}}{r_i!} =$$

$$\begin{aligned}
&= \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} t^{r_1+\dots+r_n} \frac{\left(\prod_{j=1}^n (\mu_1 p_{1j}^+ + \lambda_{01} q_{1j}) \right)^{r_1} \dots \left(\prod_{j=1}^n (\mu_n p_{nj}^+ + \lambda_{0n} q_{nj}) \right)^{r_n}}{r_1! \dots r_n!} \times \\
&\quad \times z_1^{r_1+r_2+\dots+r_n} z_2^{r_1+r_2+\dots+r_n} \dots z_n^{r_1+r_2+\dots+r_n} z_1^{-r_1} z_2^{-r_2} \dots z_n^{-r_n} = \\
&= \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} t^{r_1+\dots+r_n} \frac{\left(\prod_{j=1}^n (\mu_1 p_{1j}^+ + \lambda_{01} q_{1j}) \right)^{r_1} \dots \left(\prod_{j=1}^n (\mu_n p_{nj}^+ + \lambda_{0n} q_{nj}) \right)^{r_n}}{r_1! \dots r_n!} z_1^{R-r_1} \dots z_n^{R-r_n}, \\
& a_4(z, t) = \exp \left\{ \sum_{i,j=1}^n \mu_i p_{ij}^- q_{j0} \frac{1}{z_i z_j} t \right\} = \prod_{i=1}^n \prod_{j=1}^n \exp \left\{ \mu_i p_{ij}^- q_{j0} \frac{1}{z_i z_j} t \right\} = \\
&= \prod_{i=1}^n \prod_{j=1}^n \sum_{u_i=0}^{\infty} \frac{[\mu_i p_{ij}^- q_{j0} t z_i^{-1} z_j^{-1}]^{u_i}}{u_i!} = \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \prod_{i=1}^n \prod_{j=1}^n \frac{[\mu_i p_{ij}^- q_{j0} t z_i^{-1} z_j^{-1}]^{u_i}}{u_i!} = \\
&= \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} t^{u_1+\dots+u_n} \frac{\left(\mu_1 \prod_{j=1}^n p_{1j}^- q_{j0} \right)^{u_1} \dots \left(\mu_n \prod_{j=1}^n p_{nj}^- q_{j0} \right)^{u_n}}{u_1! \dots u_n!} \times \\
&\quad \times z_1^{-u_1-u_2-\dots-u_n} z_2^{-u_1-u_2-\dots-u_n} \dots z_n^{-u_1-u_2-\dots-u_n} z_1^{-u_1} z_2^{-u_2} \dots z_n^{-u_n} = \\
&= \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} t^{u_1+\dots+u_n} \frac{\left(\mu_1 \prod_{j=1}^n p_{1j}^- q_{j0} \right)^{u_1} \dots \left(\mu_n \prod_{j=1}^n p_{nj}^- q_{j0} \right)^{u_n}}{u_1! \dots u_n!} z_1^{-(U+u_1)} \dots z_n^{-(U+u_n)}, \\
& a_5(z, t) = \exp \left\{ \sum_{i,j=1}^n \mu_i \sum_{s=1}^n p_{ij}^- q_{js} \frac{z_s}{z_i z_j} t \right\} = \prod_{i=1}^n \prod_{j=1}^n \prod_{s=1}^n \exp \left\{ \mu_i p_{ij}^- q_{js} \frac{z_s}{z_i z_j} t \right\} = \\
&= \prod_{i=1}^n \prod_{j=1}^n \prod_{s=1}^n \sum_{w_i=0}^{\infty} \frac{[\mu_i p_{ij}^- q_{js} t z_i^{-1} z_j^{-1} z_s]^{w_i}}{w_i!} = \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} \prod_{i=1}^n \prod_{j=1}^n \prod_{s=1}^n \frac{[\mu_i p_{ij}^- q_{js} t z_i^{-1} z_j^{-1} z_s]^{w_i}}{w_i!} = \\
&= \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} t^{w_1+\dots+w_n} \frac{\left(\mu_1 \prod_{j=1}^n p_{1j}^- \prod_{s=1}^n q_{js} \right)^{w_1} \dots \left(\mu_n \prod_{j=1}^n p_{nj}^- \prod_{s=1}^n q_{js} \right)^{w_n}}{w_1! \dots w_n!} \times \\
&\quad \times z_1^{-w_1-w_2-\dots-w_n} z_2^{-w_1-w_2-\dots-w_n} \dots z_n^{-w_1-w_2-\dots-w_n} z_1^{-w_1} z_2^{-w_2} \dots z_n^{-w_n} z_1^{w_1} z_2^{w_2} \dots z_n^{w_n} = \\
&= \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} t^{w_1+\dots+w_n} \frac{\left(\mu_1 \prod_{j=1}^n p_{1j}^- \prod_{s=1}^n q_{js} \right)^{w_1} \dots \left(\mu_n \prod_{j=1}^n p_{nj}^- \prod_{s=1}^n q_{js} \right)^{w_n}}{w_1! \dots w_n!} z_1^{-W} \dots z_n^{-W}.
\end{aligned}$$

Multiplying $a_0(t)$, $a_1(z,t)$, $a_2(z,t)$, $a_3(z,t)$, $a_4(z,t)$, $a_5(z,t)$ and $\prod_{l=1}^n z_l^{\alpha_l}$ we get the expression (6).

State probability of $P(k_1, k_2, \dots, k_n, t)$ is the coefficient at $z_1^{k_1} z_2^{k_2}, \dots, z_n^{k_n}$ in the expansion of function of $\Psi_n(z, t)$ in a multiple series of (8), on condition that at the initial time the network is in a state of $(\alpha_1, \alpha_2, \dots, \alpha_n, 0)$.

3. Finding the average number of messages at network systems

It is known that the expectation of m -th components of a multi-dimensional random variable can be found by differentiating the generating function (6) by z_m and assumed $z_i = 1, i = 1, n$. Therefore, the average number of messages in the system S_m of network we will use the relation:

$$\begin{aligned} \frac{\partial \Psi_n(z, t)}{\partial z_m} &= \sum_{l_1=0}^{\infty} \dots \sum_{l_n=0}^{\infty} \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \\ &\sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} (\alpha_m + l_m - c_m - r_m + R - u_m - U - W) t^{\sum_{i=1}^n (l_i + c_i + r_i + u_i + w_i)} \times \\ &\times \prod_{i=1}^n \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{c_i} \prod_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})^{r_i} \left(\mu_i \prod_{j=1}^n p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^n \prod_{s=1}^n q_{js} \right)^{w_i}}{l_i! c_i! r_i! u_i! w_i!} \times \\ &\times z_m^{\alpha_m + l_m - c_m - r_m + R - u_m - U - W - 1} \prod_{\substack{i=1, \\ i \neq m}}^n z_i^{\alpha_i + l_i - c_i - r_i + R - u_i - U - W}. \end{aligned}$$

It follows that the average number of messages in the system S_m , is defined by the formula

$$\begin{aligned} N_m(t) &= \left. \frac{\partial \Psi_n(z, t)}{\partial z_m} \right|_{z=(1,1,\dots,1)} = a_0(t) \sum_{l_1=0}^{\infty} \dots \sum_{l_n=0}^{\infty} \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \\ &\sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} (\alpha_m + l_m - c_m - r_m + R - u_m - U - W) t^{\sum_{i=1}^n (l_i + c_i + r_i + u_i + w_i)} \times \quad (7) \\ &\times \prod_{i=1}^n \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{c_i} \prod_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})^{r_i} \left(\mu_i \prod_{j=1}^n p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^n \prod_{s=1}^n q_{js} \right)^{w_i}}{l_i! c_i! r_i! u_i! w_i!}. \end{aligned}$$

We make the expression (7) the change of variables

$$k_m = \alpha_m + l_m - c_m - r_m + R - u_m - U - W,$$

then $l_m = k_m - \alpha_m + c_m + r_m - R + u_m + U + W$ and

$$\begin{aligned} N_m(t) = & a_0(t) \sum_{c_1=0}^{\infty} \dots \sum_{c_n=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \\ & \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} \sum_{k_1=\alpha_1-c_1-r_1+R-u_1-U-W}^{\infty} \dots \sum_{k_n=\alpha_n-c_n-r_n+R-u_n-U-W}^{\infty} k_m \times \\ & \sum_{t=1}^n (k_i - \alpha_i + 2c_i + 2r_i - R + 2u_i + U + W) \times \\ & \times \prod_{i=1}^n \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{c_i} \prod_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})^{r_i} \left(\mu_i \prod_{j=1}^n p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^n \prod_{s=1}^n q_{js} \right)^{w_i}}{(k_i - \alpha_i + c_i + r_i - R + u_i + U + W)! q_i! r_i! u_i! w_i!}. \end{aligned}$$

Because network systems operate under high load, then

$$k_i = \alpha_i - c_i - r_i + R - u_i - U - W \geq 1$$

and, therefore, $c_i \leq \alpha_i - r_i + R - u_i - U - W - 1$, so

$$\begin{aligned} N_m(t) = & a_0(t) \sum_{k_1=0}^{\infty} \dots \sum_{k_m=0}^{\infty} k_m \dots \sum_{k_n=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_n=0}^{\infty} \\ & \sum_{u_1=0}^{\infty} \dots \sum_{u_n=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_n=0}^{\infty} \sum_{c_1=0}^{\alpha_1-r_1+R-u_1-U-W-1} \dots \sum_{c_n=0}^{\alpha_n-r_n+R-u_n-U-W-1} \sum_{t=1}^n (k_i - \alpha_i + 2c_i + 2r_i - R + 2u_i + U + W) \times (8) \\ & \times \prod_{i=1}^n \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{c_i} \prod_{j=1}^n (\mu_i p_{ij}^+ + \lambda_{0i}^- q_{ij})^{r_i} \left(\mu_i \prod_{j=1}^n p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^n \prod_{s=1}^n q_{js} \right)^{w_i}}{(k_i - \alpha_i + c_i + r_i - R + u_i + U + W)! c_i! r_i! u_i! w_i!}. \end{aligned}$$

To find state probabilities and the average number of messages in a network systems a program was written in a mathematical computation package Mathematica.

4. Examples

Let the number of QS in the network $n = 3$. The intensity of the input stream of positive and signals λ_{0i}^+ and λ_{0i}^- are equal respectively to: $\lambda_{01}^+ = 1/2$, $\lambda_{02}^+ = 1/5$,

$\lambda_{03}^+ = 1/3$, $\lambda_{01}^- = 1/3$, $\lambda_{02}^- = 1/4$, $\lambda_{03}^- = 1/6$. The intensities of message service μ_i equal: $\mu_1 = 1/2$, $\mu_2 = 4$, $\mu_3 = 1/3$. We assume, that the transition probability of positive messages p_{ij}^+ have form: $p_{12}^+ = 1/9$, $p_{13}^+ = 2/7$, $p_{21}^+ = 1/8$, $p_{23}^+ = 1/5$, $p_{31}^+ = 2/9$, $p_{32}^+ = 1/7$; the transition probability of signals p_{ij}^- equals: $p_{12}^- = 1/8$, $p_{13}^- = 1/7$, $p_{21}^- = 3/7$, $p_{23}^- = 1/9$, $p_{31}^- = 2/7$, $p_{32}^- = 2/11$; output probability from network of messages to the external environment p_{i0} equals $p_{10} = 169/504$, $p_{20} = 341/2520$, $p_{30} = 116/693$. Suppose also, that the probability that the incoming signal immediately moves the positive message from the system S_j to system S_s equal $q_{12} = 1/3$, $q_{13} = 1/3$, $q_{21} = 1/4$, $q_{23} = 1/5$, $q_{31} = 1/3$, $q_{32} = 1/7$. The signal is triggered by a negative message and is destroyed at the QS S_j positive message with probability $q_{10} = 1/3$, $q_{20} = 11/20$, $q_{30} = 11/21$. Expression (7) in this case is equal to $a_0(t) = e^{-\frac{397t}{60}}$.

For example, we need to find the probability of the state $P(3,3,3,t)$. It is the coefficient of $z_1^3 z_2^3 z_3^3$ in the expansion of $\Psi_n(z,t)$ in multiple series (6), so the degrees at z_i must satisfy the relation $\alpha_i + l_i - c_i - r_i + R - u_i - U - W = 3$, $i = \overline{1,3}$, this implies that

$$c_i = \alpha_i + l_i + \sum_{\substack{j=1 \\ j \neq i}}^3 r_j - \sum_{\substack{j=1 \\ j \neq i}}^3 u_j - \sum_{j=1}^3 w_j - 3,$$

$$\sum_{i=1}^3 (l_i + c_i + r_i + u_i + w_i) = \sum_{i=1}^3 (\alpha_i + 2l_i) + 3(R - U - W - 3), \quad i = \overline{1,3}.$$

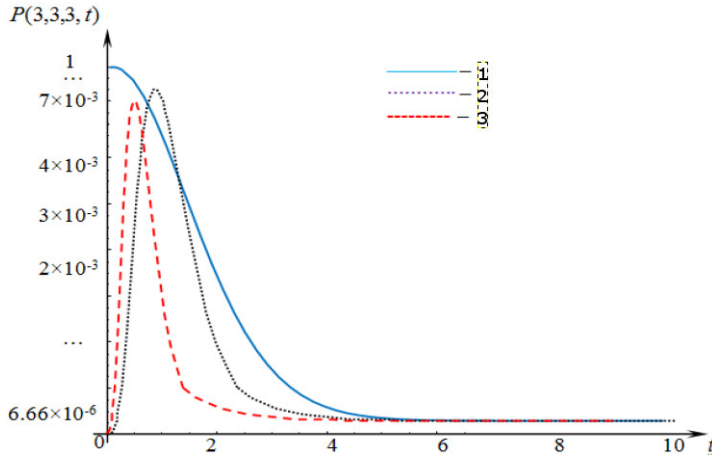
Then from (6) we obtain that

$$P(3,3,3,t) = e^{-\frac{397t}{60}} \sum_{l_1=0}^{\infty} \dots \sum_{l_3=0}^{\infty} \sum_{r_1=0}^{\infty} \dots \sum_{r_3=0}^{\infty} \sum_{u_1=0}^{\infty} \dots \sum_{u_3=0}^{\infty} \sum_{w_1=0}^{\infty} \dots \sum_{w_3=0}^{\infty} t^{i=1}^{\sum (\alpha_i + 2l_i) + 3(R - U - W - 3)} \times$$

$$\times \prod_{i=1}^3 \frac{\lambda_{0i}^+ l_i (\mu_i p_{i0} + \lambda_{0i}^- q_{i0})^{\alpha_i + l_i + R - U - u_i - W - 10}}{l_i! (\alpha_i + l_i + R - U - u_i - W - 3)! r_i! u_i! w_i!} \times$$

$$\times \prod_{j=1}^3 (\mu_j p_{ij}^+ + \lambda_{0j}^- q_{ij})^{r_j} \left(\mu_i \prod_{j=1}^3 p_{ij}^- q_{j0} \right)^{u_i + w_i} \left(\prod_{j=1}^3 \prod_{s=1}^3 q_{js} \right)^{w_i}, \quad \alpha_i = 3, i = \overline{1,3}.$$

Figure 1 shows plots of the state probability of $P(3,3,3,t)$ and in condition that, at the initial time moment network is in one of three states: 1) $\alpha_i = 3$, $i = \overline{1,3}$, 2) $\alpha_i = 1$, $i = \overline{1,3}$, 3) $\alpha_i = 0$, $i = \overline{1,3}$.

Fig. 1. Graph of state probability $P(3,3,3,t)$

We find the state probabilities and the average number of messages in network systems at the stationary behavior. For G-networks with negative messages and triggers the stationary distribution $P(k)$ appears in the multiplicative form [3, 7]:

$$P(k) = \prod_{i=1}^n p(k_i), \quad p(k_i) = (1 - q_i) q_i^{k_i}, \quad q_i < 1, \quad k_i \geq 0,$$

where

$$q_i = \frac{\lambda_i^+}{\lambda_i^- (1 - q_{i0}) + \lambda_i^- q_{i0} + \mu_i},$$

$$\lambda_i^+ = \lambda_{0i}^+ + \sum_{j=1}^n q_j \mu_j \left[p_{ji}^+ + \sum_{s=1}^n p_{js}^- q_s q_{si} \right] + \sum_{j=1}^n \lambda_{0j}^- q_j q_{ji},$$

$$\lambda_i^- = \lambda_{0i}^- + \sum_{j=1}^n q_j \mu_j p_{ji}^-, \quad i = \overline{1, n}.$$

The average number of messages located in system S_m , is calculated as follows:

$$N_m = \sum_{k_1=0}^{\infty} \sum_{k_2=0}^{\infty} \dots \sum_{k_m=0}^{\infty} k_m \dots \sum_{k_n=0}^{\infty} P(k), \quad m = \overline{1, n}.$$

The calculations in the package Mathematica showed that the stationary probabilities of the above conditions are equal to respectively $P(3,3,3) = 6.66 \times 10^{-6}$, $P(1,1,1) = 0,003$, $P(0,0,0) = 0,07$, and the average number messages is equal to respectively $N_1 = 0.0004$, $N_3 = 0.0003$, $N_3 = 0.0006$.

The average number of requests in the system network (in the queue and on service) on condition that $N_m(0) = 0$, $m = 1, n$, can be found by the formula (8).

Figure 2 shows a graph of the average number of messages in the QS S_1 in the network form example above.

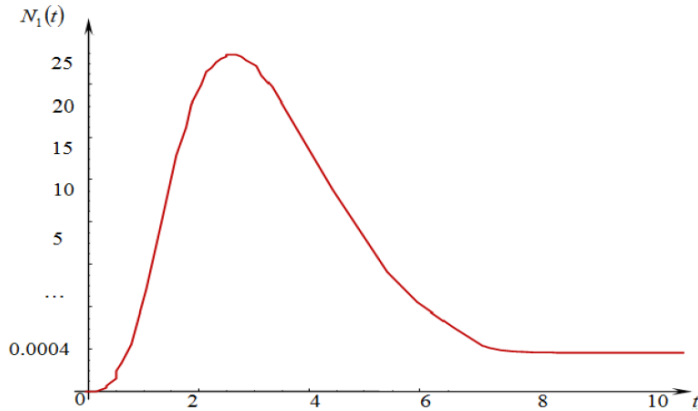


Fig. 2. Graph change in the average number of messages $N_1(t)$ in QS S_1

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