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**Determining reliability  
parameters for a commodity  
transfer system with non-  
independent components**

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## 1. Introduction

A transmission network composed of  $n$  linearly arranged components  $e_0, \dots, e_n$  is considered. For each  $i=1, \dots, n-1$  the component  $e_i$  is directly connected to  $e_{i-1}$  and  $e_{i+1}$ , while  $e_n$  is directly connected to  $e_{n-1}$  alone;  $e_0$  is the source component from which certain commodity (electric power, radio signal, electronic data, water, gas, etc.) is transferred, via  $e_1, \dots, e_{n-1}$ , to  $e_n$ . Each component can be in one of two states: 1 – operating, 0 – failed;  $e_0$  is always in operating state. The repair of a failed component is started as soon as one of repair teams is available – due to a limited number of them the repair may not start immediately after the component's failure. The order in which failed components are chosen for repair depends on the repair policy applied – two of such policies will be considered. The time-to-failure and time-to-repair of  $e_i$  are random variables with distribution functions  $F_i$  and  $G_i$  respectively.

A commodity can be transferred from  $e_0$  to  $e_i$ ,  $i=1, \dots, n$ , if and only if  $e_i$  is functional and connected to  $e_0$ , i.e.  $e_1, \dots, e_i$  are in the operating state. As failures of components occur, the periods during which functional  $e_i$  is connected to  $e_0$  are interleaved by the periods during which  $e_i$  is failed or disconnected from  $e_0$ . The main goal of this paper is to determine the mean durations of these time intervals, i.e. the mean time from the moment when the connection between  $e_0$  and  $e_i$  is interrupted to the moment when it is restored, and the mean time of uninterrupted connection between  $e_0$  and  $e_i$ .

Most probabilistic models of complex systems assume full independence of their components. In particular, it means that the components' lifetimes and repair times are independent random variables. Obviously, this assumption is made for the sake of computational simplicity. In reality, however, such independence rarely occurs. In order to make the considered network model more true-to-life it is assumed that the functioning of  $e_i$  depends on the states of  $e_1, \dots, e_{i-1}$  in the following way:  $e_i$  can only fail if  $e_1, \dots, e_{i-1}$  are in the operating state; as long as this condition is fulfilled the time-to-failure of  $e_i$  has the distribution function  $F_i$ . In consequence, an element cannot

fail if it is disconnected from  $e_0$ . This conveys the idea that only components being “under load” are failure prone as is the case in many real-life systems. Thus, it can be said that  $e_i$  functions independently of  $e_1, \dots, e_{i-1}$  up to the moment when one of  $e_1, \dots, e_{i-1}$  fails. It must be stressed that  $e_i$  functions independently of  $e_{i+1}, \dots, e_n$ , moreover, a component's repair time is independent on the behavior of other components.

It is clear that the network's behavior and, consequently, mean connectedness and non-connectedness times of components depend on two factors: the number of repair teams assigned to the network maintenance, and the repair policy implemented. As to the first factor, the greater the number of repair teams, the shorter the mean non-connectedness times of individual components, and vice versa. If there are fewer than  $n$  repair teams, a component's repair may not start immediately after its failure – the average time of delay decreases with the total number of repair teams and is equal to zero if this number reaches  $n$ . However, this case should be considered only for theoretical purposes, because in practice the number of repair teams is usually considerably smaller than the number of system's components. As follows from the above argument, the mean non-connectedness times of components in the cases of one and  $n$  repair teams are the upper and lower bounds of these mean times in each remaining case (more than one and less than  $n$  repair teams).

Passing to the subject of repair policies, two of them will be considered. According to the first policy, the components are chosen for repair in the same order in which they fail. As only the components connected to  $e_0$  can fail, the next component selected for repair (by the first available team) is the one farthest from  $e_0$ . This means that the queue of components waiting to be repaired is of the FIFO type. The second policy consists in prioritizing the components which are least distant from  $e_0$ , i.e. the next component selected for repair is the one nearest to  $e_0$ . Thus the order in which failed components are chosen for repair is reverse to the order in which they fail. This means that the queue of components waiting to be repaired is of the LIFO type.

Obviously, choosing between multiple maintenance policies makes sense only if there are less than  $n$  repair teams. Otherwise the only feasible policy to follow is “repair a component upon its failure”, as there is always at least one team available when a component fails.

## 2. Definitions and notation

Throughout the paper the following notation will be used:

$L_i$  – time-to-failure for  $e_i$ , provided that  $e_1, \dots, e_{i-1}$  remain in operational state up to the failure of  $e_i$

$R_i$  – time-to-repair for  $e_i$

$F_i, G_i$  – distribution functions of  $L_i$  and  $G_i$  respectively;

$\varphi_j^{(i)}$  – time of the  $j$ -th disconnection of  $e_i$  from  $e_0$ ,  $j \geq 1$ ;

$\rho_j^{(i)}$  – time of the  $j$ -th reconnection of  $e_i$  to  $e_0$ ,  $j \geq 1$ ; it is assumed that  $\rho_0^{(i)} = 0$ ;

$\chi_j^{(i)}$  – length of time from  $\rho_{j-1}^{(i)}$  to  $\varphi_j^{(i)}$ ,  $j \geq 1$ , i.e.  $\chi_j^{(i)} = \varphi_j^{(i)} - \rho_{j-1}^{(i)}$  is the length of the  $j$ -th period during which  $e_i$  remains connected to  $e_0$

$\psi_j^{(i)}$  – length of time from  $\varphi_j^{(i)}$  to  $\rho_j^{(i)}$ ,  $j \geq 1$ , i.e.  $\psi_j^{(i)} = \rho_j^{(i)} - \varphi_j^{(i)}$  is the length of the  $j$ -th period during which  $e_i$  remains disconnected from  $e_0$

$\pi_1$  – the “first failed, first selected” policy;

$\pi_2$  – the “last failed, first selected” policy;

$C_j(i, n, r, s), D_j(i, n, r, s)$  – the expected values of  $\chi_j^{(i)}$  and  $\psi_j^{(i)}$ , provided that the system is composed of  $n$  components, the number of repair teams equals  $r$ , and the repair policy  $\pi_s$  is applied,  $s = 1, 2$ ;

$C(i, n, r, s), D(i, n, r, s)$  – the expected values of limiting means of  $\chi_j^{(i)}$  and  $\psi_j^{(i)}$ , provided the limiting means exist.

We thus have:

$$\begin{aligned}
 C_j(i, n, r, s) &= E(\chi_j^{(i)}) \\
 D_j(i, n, r, s) &= E(\psi_j^{(i)}) \\
 C(i, n, r, s) &= E[\lim_{n \rightarrow \infty} \sum_{j=1}^n \chi_j(i, n, r, s)] \\
 D(i, n, r, s) &= E[\lim_{n \rightarrow \infty} \sum_{j=1}^n \psi_j(i, n, r, s)]
 \end{aligned}$$

### 3. Analytical computation of $C(i, n, r, s)$ and $D(i, n, r, s)$ for a two-component system

In this chapter the two basic reliability parameters, i.e.  $C(i, n, r, s)$  and  $D(i, n, r, s)$  will be computed analytically for a two-component system. The additional assumption, making the computations possible, is that  $L_1, L_2, R_1, R_2$  are exponentially distributed, i.e.

$$(1) \quad F_i(t) = 1 - \exp(-\lambda_i t), \quad G_i(t) = 1 - \exp(-\mu_i t), \quad i = 1, 2$$

The definitions given in chapter 2 imply that every period of time during which both  $e_1$  and  $e_2$  remain connected to  $e_0$  falls within the limits of one of the intervals  $[\rho_{j-1}^{(2)}, \varphi_j^{(2)}], j \geq 1$ , while every period of time when  $e_1$  or  $e_2$  remains disconnected from  $e_0$  falls within the limits of one of the intervals  $[\varphi_j^{(2)}, \rho_j^{(2)}], j \geq 1$ . As the lifetime distributions of  $e_1$  and  $e_2$  are exponential, the system's behavior is stochastically identical on each interval  $[\rho_{j-1}^{(2)}, \rho_j^{(2)}], j \geq 1$ . Thus, by the end of this chapter, let time be measured from any moment  $\rho_j^{(2)}, j \geq 0$ , i.e. from the moment 0 or any other moment when  $e_2$  is reconnected to  $e_0$ .

It is very easy to compute  $C_j(i, n, r, s)$ . Indeed, for any  $j \geq 1$  the components  $e_1, \dots, e_i$  are all operational at the instant  $\rho_{j-1}^{(i)}$ , and the time from  $\rho_{j-1}^{(i)}$  to the failure of  $e_k$  is exponentially distributed,  $k=1, \dots, i$  (the "lack of memory" property of the exponential distribution). We thus have:

$$(2) \quad P(\mathcal{X}_j^{(i)} \leq t) = P(\min(L_1, \dots, L_i) \leq t) = 1 - \exp[-(\lambda_1 + \dots + \lambda_i)]$$

From (2) it follows that

$$(3) \quad C_j(i) = C(i) = \frac{1}{\lambda_1 + \dots + \lambda_i}, \quad j \geq 1$$

The analytical computation of  $D_j(i, n, r, s)$  is more difficult, however it will be presented for  $n=2$ . It should be noted that for a two-component system both repair policies are equivalent. Indeed, if  $r=1$  then the queue of components awaiting repair has maximum length 1, if  $r=2$  then the queue always has zero length. In both cases no selection decision has to be made. First, the case of one repair team will be considered. Let the events A, B, and C be defined as follows:

$$(4) \quad \begin{aligned} A &= \{L_1 < L_2\} \\ B &= \{L_1 > L_2, R_2 > L_1 - L_2\} \\ C &= \{L_1 > L_2, R_2 < L_1 - L_2\} \end{aligned}$$

Clearly, A, B, and C form a complete system of events. As  $e_1$  can be disconnected from  $e_0$  only in case of the event A or B, the expected time-to-reconnection for  $e_1$ , provided that  $e_1$  was disconnected from  $e_0$ , is equal to  $E(\psi_j^{(1)} | A \cup B)$ . The latter symbol denotes the conditional expectation of  $\psi_j^{(1)}$ , given the event  $A \cup B$ .

In general, the conditional expectation of a random variable X, given the event E such that  $P(E) > 0$ , is defined as follows:

$$(5) \quad \int_{-\infty}^{+\infty} t dF_{(X|E)}(t),$$

where

$$(6) \quad F_{(X|E)} = \frac{1}{P(E)} P(\{X \leq t\} \cap E)$$

$F_{(X|E)}$  is called the conditional distribution function of  $X$ , given the event  $E$ .

Following the above definition, we obtain

$$(7) \quad E(\psi_j^{(1)} | A \cup B) = \frac{E(\psi_j^{(1)} | A)P(A) + E(\psi_j^{(1)} | B)P(B)}{P(A) + P(B)}$$

In case of the event  $A$ , i.e. when  $e_1$  fails before  $e_2$ , the repair of  $e_1$  starts immediately after its failure, and  $e_2$  cannot fail until the repair of  $e_1$  is completed. The time-to-reconnection for  $e_1$  is thus equal to  $R_1$ , therefore

$$(8) \quad E(\psi_j^{(1)} | A) = E(R_1) = \frac{1}{\mu_1}$$

with

$$(9) \quad P(A) = \int_0^{\infty} P(L_2 > x) dF_1(x) = \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

In case of the event  $B$  the failure of  $e_2$  precedes that of  $e_1$ , while the repair of  $e_2$ , starting immediately after its failure, ends after the failure of  $e_1$ , being followed by the repair of  $e_1$ . We thus have:



$$(10) \quad E(\psi_j^{(1)} | B) = E(R_2^{res} | B) + E(R_1)$$

where  $R_2^{res}$  is equal to the residual repair time of  $e_2$ , i.e. the time elapsed from the failure of  $e_1$  to the completion of  $e_2$ 's repair. It is also true that

$$(11) \quad P(\{R_2^{res} \leq t\} \cap B) = \int_0^{\infty} \int_0^x \Pr(x - y < R_2 \leq x - y + t) dF_2(y) dF_1(x)$$

and

$$(12) \quad P(B) = \int_0^{\infty} \int_0^x \Pr(R_2 > x - y) dF_2(y) dF_1(x) = \frac{\lambda_2}{(\lambda_1 + \lambda_2)} \cdot \frac{\lambda_1}{(\lambda_1 + \mu_2)}$$

As

$$(13) \quad \Pr(x - y < R_2 \leq x - y + t) = \exp[-\mu_2(x - y)][1 - \exp(-\mu_2 t)] = G_2(t) \Pr(R_2 > x - y),$$

the following equality holds:

$$(14) \quad P(\{R_2^{res} \leq t\} | B) = G_2(t)$$

meaning that the conditional distribution function of  $R_2^{res}$ , given the event B, is equal to the distribution function of  $R_2$ . In view of (10) this result yields:

$$(15) \quad E(\psi_j^{(1)} | B) = E(R_2) + E(R_1) = \frac{1}{\mu_1} + \frac{1}{\mu_2}$$

Summing up, we obtain the following formula:

$$(16) \quad D(1,2,1) = D_j(1,2,1) = \frac{1}{\mu_1} + \frac{1}{\mu_2} \cdot \frac{\lambda_2}{\lambda_1 + \lambda_2 + \mu_2}$$

Based on a similar argument as in the case of  $e_1$ , the expected time-to-reconnection for  $e_2$ , provided that  $e_2$  was disconnected from  $e_0$ , is equal to  $E(\psi_j^{(2)} | A \cup B \cup C)$  – each of the events A, B, and C results in disconnecting  $e_2$  from  $e_0$ . Using the total probability law, we obtain:

$$(17) \quad E(\psi_j^{(2)} | A \cup B \cup C) = E(\psi_j^{(2)} | A)P(A) + E(\psi_j^{(2)} | B)P(B) + E(\psi_j^{(2)} | C)P(C)$$

In case of the event A, the time-to-reconnection for  $e_2$  is equal to that of  $e_1$ . We thus have:

$$(18) \quad E(\psi_j^{(2)} | A) = E(\psi_j^{(1)} | A) = \frac{1}{\mu_1}$$

In case of the event B, the time-to-reconnection for  $e_2$  is equal to  $R_2 + R_1$ , therefore

$$(19) \quad E(\psi_j^{(2)} | B) = E(R_2) + E(R_1) = \frac{1}{\mu_1} + \frac{1}{\mu_2}$$

In case of the event C, the repair of  $e_2$  starts and finishes before the failure of  $e_1$ , hence

$$(20) \quad E(\psi_j^{(2)} | C) = E(R_2) = \frac{1}{\mu_2}$$

with

$$(21) \quad P(C) = \int_0^{\infty} \int_0^x \Pr(R_2 < x - y) dF_2(y) dF_1(x) = \frac{\lambda_2}{(\lambda_1 + \lambda_2)} \cdot \frac{\mu_2}{(\lambda_1 + \mu_2)}$$

Finally the following result is obtained:

$$(22) \quad D(2,2,1) = D_j(2,2,1) = \frac{1}{\mu_1} \cdot \frac{\lambda_1}{\lambda_1 + \lambda_2} + \left( \frac{1}{\mu_1} + \frac{1}{\mu_2} \right) \cdot \frac{\lambda_2}{(\lambda_1 + \lambda_2)} \cdot \frac{\lambda_1}{(\lambda_1 + \mu_2)} + \frac{1}{\mu_2} \cdot \frac{\lambda_2}{(\lambda_1 + \lambda_2)} \cdot \frac{\mu_2}{(\lambda_1 + \mu_2)}$$

Now the case of two repair teams will be considered. As  $e_1$  fails independently of the state of  $e_2$ , and one repair team is always available for  $e_1$ , the failure and repair process of  $e_1$  is an alternating renewal process independent of the state of  $e_2$ . The mean time-to-reconnection for  $e_1$  is thus given by  $E(R_1)$ , i.e:

$$(23) \quad D(1,2,2) = D_j(1,2,2) = \frac{1}{\mu_1}$$

Let  $\{Y(t), t \geq 0\}$  denote the failure and repair process of  $e_1$ , where  $Y(t)=1$  if  $e_1$  is operational at the time  $t$ , otherwise ( $e_1$  is under repair)  $Y(t)=0$ . Let  $Z(t)$  denote the time elapsing from  $t$  to the next state change of  $e_1$ . Obviously,  $\{Z(t), t > 0\}$  is also a stochastic process. As  $L_1$  and  $R_1$  are exponentially distributed, for the process  $Z$  we have:

$$(24) \quad \begin{aligned} \Pr(Z(t) \leq s \mid Y(t) = 1) &= 1 - \exp(-\lambda_1 s) \\ \Pr(Z(t) \leq s \mid Y(t) = 0) &= 1 - \exp(-\mu_1 s) \end{aligned}$$

and for the process Y:

$$(25) \quad \begin{aligned} \Pr(Y(t+s) = 1 \mid Y(t) = 1) &= \frac{\mu_1}{\lambda_1 + \mu_1} + \frac{\lambda_1}{\lambda_1 + \mu_1} \exp[-(\lambda_1 + \mu_1)s] \\ \Pr(Y(t+s) = 0 \mid Y(t) = 1) &= \frac{\lambda_1}{\lambda_1 + \mu_1} [1 - \exp[-(\lambda_1 + \mu_1)s]] \end{aligned}$$

The latter are the formulas for the transition probabilities of Y which is a Markov process (see [6]).

Let the events A, B and C be defined in the following way:

$$(26) \quad \begin{aligned} A &= \{Z(0) < L_2\} \\ B &= \{Z(0) > L_2, \quad Y(L_2 + R_2) = 0\} \\ C &= \{Z(0) > L_2, \quad Y(L_2 + R_2) = 1\} \end{aligned}$$

Clearly, A, B, and C form a complete system of events. As  $e_2$  is disconnected from  $e_0$  in case of the event A, B or C, the expected time-to-reconnection for  $e_2$ , provided that  $e_2$  has been disconnected from  $e_0$ , is equal to  $E(\psi_j^{(2)} \mid A \cup B \cup C)$ . Using the total probability law, we obtain:

$$(27) \quad D(2,2,2) = D_j(2,2,2) = E(\psi_j^{(2)} \mid A)P(A) + E(\psi_j^{(2)} \mid B)P(B) + E(\psi_j^{(2)} \mid C)P(C)$$

In the case of A, the time-to-reconnection for  $e_2$  is equal to that of  $e_1$ . We thus have:

$$(28) \quad E(\psi_j^{(2)} \mid A) = \frac{1}{\mu_1}$$

with

$$(29) \quad P(A) = \int_0^{\infty} P(L_2 > x) dF_1(x) = \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

In case of the event B the repair of  $e_1$ , starting before the failure of  $e_2$ , continues for some time after the repair of  $e_2$  has been completed. We thus have:

$$(30) \quad E(\psi_j^{(2)} | B) = E(R_2) + E(R_1^{\text{res}} | B)$$

where  $R_1^{\text{res}}$  is equal to the residual repair time of  $e_1$ , i.e. the time elapsed from the completion of  $e_1$ 's repair to the completion of  $e_2$ 's repair. Using (24) we obtain:

$$\begin{aligned} P(\{R_1^{\text{res}} \leq t\} \cap B) &= \{Z(0) > L_2, Y(L_2 + R_2) = 0, Z(L_2 + R_2) \leq t\} = \\ &= \int_0^{\infty} P(Z(x+y) \leq t, Y(x+y) = 0, Z(0) > x) dG_2(y) dF_2(x) = \\ (31) \quad &= \int_0^{\infty} P(Z(x+y) \leq t | Y(x+y) = 0) \times \\ &\quad \times P(Y(x+y) = 0, Z(0) > x) dG_2(y) dF_2(x) = \\ &= \exp(-\mu_1 t) \int_0^{\infty} P(Y(x+y) = 0, Z(0) > x) dG_2(y) dF_2(x) = \\ &= \exp(-\mu_1 t) P(B) \end{aligned}$$

while (25) yields:

$$\begin{aligned}
P(B) &= \int_0^{\infty} P(Y(x+y) = 0 \mid Z(0) > x) P(Z(0) > x) dG_2(y) dF_2(x) = \\
&= \int_0^{\infty} P(Y(x+y) = 0 \mid Y(x) = 1) P(Z(0) > x) dG_2(y) dF_2(x) = \\
(32) \quad &= \int_0^{\infty} \frac{\lambda_1}{\lambda_1 + \mu_1} [1 - \exp(-\lambda_1 y - \mu_1 y)] \exp(-\lambda_1 x) dG_2(y) dF_2(x) = \\
&= \frac{\lambda_2}{\lambda_1 + \lambda_2} \cdot \frac{\lambda_1}{\lambda_1 + \mu_1} \cdot \left( 1 - \frac{\mu_2}{\lambda_1 + \mu_1 + \mu_2} \right)
\end{aligned}$$

From (31) it follows that:

$$(33) \quad P(\{R_1^{\text{res}} \leq t\} \mid B) = G_1(t)$$

meaning that the conditional distribution function of  $R_1^{\text{res}}$ , given the event B, is equal to the distribution function of  $R_1$ . In view of (30) this result yields:

$$(34) \quad E(\psi_j^{(2)} \mid B) = E(R_2) + E(R_1) = \frac{1}{\mu_1} + \frac{1}{\mu_2}$$

In case of the event C,  $e_1$  is operational when the repair of  $e_2$  ends, thus the time-to-reconnection for  $e_2$  is equal to  $R_2$ , which means that

$$(35) \quad E(\psi_j^{(2)} \mid C) = E(R_2) = \frac{1}{\mu_2}$$

where, using (25),  $P(C)$  is computed as follows:

$$\begin{aligned}
P(C) &= \int_0^{\infty} P(Y(x+y) = 1 \mid Z(0) > x) P(Z(0) > x) dG_2(y) dF_2(x) = \\
&= \int_0^{\infty} P(Y(x+y) = 1 \mid Y(x) = 1) P(Z(0) > x) dG_2(y) dF_2(x) = \\
(36) \quad &= \int_0^{\infty} \left[ \frac{\mu_1}{\lambda_1 + \mu_1} + \frac{\lambda_1}{\lambda_1 + \mu_1} \exp(-\lambda_1 y - \mu_1 y) \right] \exp(-\lambda_1 x) dG_2(y) dF_2(x) = \\
&= \frac{\lambda_2}{\lambda_1 + \lambda_2} \left( \frac{\mu_1}{\lambda_1 + \mu_1} + \frac{\lambda_1}{\lambda_1 + \mu_1} \cdot \frac{\mu_2}{\lambda_1 + \mu_1 + \mu_2} \right)
\end{aligned}$$

Finally the following result is obtained:

$$\begin{aligned}
D(2,2,2) &= D_j(2,2,2) = \frac{1}{\mu_1} \cdot \frac{\lambda_1}{\lambda_1 + \lambda_2} + \\
(37) \quad &+ \left( \frac{1}{\mu_1} + \frac{1}{\mu_2} \right) \cdot \frac{\lambda_2}{\lambda_1 + \lambda_2} \cdot \frac{\lambda_1}{\lambda_1 + \mu_1} \cdot \left( 1 - \frac{\mu_2}{\lambda_1 + \mu_1 + \mu_2} \right) + \\
&+ \frac{1}{\mu_2} \cdot \frac{\lambda_2}{\lambda_1 + \lambda_2} \cdot \left( \frac{\mu_1}{\lambda_1 + \mu_1} + \frac{\lambda_1}{\lambda_1 + \mu_1} \cdot \frac{\mu_2}{\lambda_1 + \mu_1 + \mu_2} \right)
\end{aligned}$$

In conclusion, one remark should be made. For the system under consideration, the analytical method seems to be of minor practical significance, mainly due to the assumption that random variables describing components' behavior are exponentially distributed, but also because of enormously complex formulas that would be derived for  $n > 2$ . It should be underscored that analytical computation of  $D(i,n,r,s)$  for large  $n$  is an open problem that can possibly be solved using some recursive method. Nevertheless, the results obtained for  $n=2$  can be helpful in testing the correctness and accuracy of the simulation method presented in the next chapter.

#### 4. Computing $C(i,n,r,s)$ and $D(i,n,r,s)$ by means of Monte Carlo Simulation

In this chapter algorithms for estimating  $C(i,n,r,s)$  and  $D(i,n,r,s)$ , based on Monte Carlo simulation are presented. When estimating the parameters of a stochastic process one often encounters the following problem: whether statistical data may come from one sample path (realization) of the process or should they be collected from multiple sample paths? Clearly, one sample path is sufficient in the case of a recurrent process, i.e. a process  $X=\{X(t), t \geq 0\}$  with the following properties:

- the state of  $X$  at  $t = 0$  is fixed, i.e.  $X(0)$  has the one-point distribution,
- with probability one  $X$  returns to the state  $X(0)$  after finite time, measured from 0,
- if  $Y(t)=X(\tau_1+t)$ , where  $\tau_1$  is the (random) time of the first return of  $X$  to its initial state, then  $Y=\{Y(t), t \geq 0\}$  and  $X$  are stochastically identical processes ( $X$  begins anew at  $t = \tau_1$ ).

Sometimes the second property is replaced with the stronger one, i.e.  $E(\tau_1) < \infty$ . For details see [5].

##### Lemma 1

If the components' lifetimes (the random variables  $L_1, \dots, L_n$ ) are exponentially distributed, and the repair policy  $\pi_1$  is applied, then the vector valued process  $\underline{X}=\{[X_1(t), \dots, X_n(t)], t \geq 0\}$ , where  $X_i(t)$  denotes the state of  $e_i$  at time  $t$ , is recurrent.

##### Proof

Note that  $\underline{X}(0)=\underline{1}$  and  $\underline{X}$  begins anew at any moment  $t$  when  $\underline{X}(t)=\underline{1}$ , due to the "lack of memory" property of the exponential distribution. Moreover,

$$(38) \quad \tau_1 = \rho_1^{(n)} = \chi_1^{(n)} + \psi_1^{(n)} \leq \min(L_1, \dots, L_n) + R_1 + \dots + R_n,$$

The above inequality becomes equality only in the "worst" case, i.e.  $r=1$ ,  $e_n$  is the first failed



component, and  $e_{i-1}$  fails before  $e_i$  has been repaired,  $2 \leq i \leq n$ . Otherwise we have strong inequality in (38), provided that  $\Pr(R_i > 0) = 1$ ,  $1 \leq i \leq n$ . From (38) it follows that

$$(39) \quad E(\tau_1) \leq \frac{1}{(\lambda_1 + \dots + \lambda_n)} + \sum_{i=1}^n E(R_i),$$

thus the stronger version of the second property is fulfilled if  $E(R_i) < \infty$ ,  $1 \leq i \leq n$ . For the weaker version it is sufficient that  $\Pr(R_i < \infty) = 1$ .

If the policy  $\pi_2$  is applied, then the question whether  $\underline{X}$  is recurrent remains open, even for exponentially distributed  $L_1, \dots, L_n$ . Most likely, some additional assumptions regarding the distribution functions  $G_1, \dots, G_n$  should be made to ensure that  $\underline{X}$  has this property.

It follows from Lemma 1 that  $\chi_1, \chi_2, \dots$  are independent identically distributed random variables (IIDRV). However,  $\psi_1, \psi_2, \dots$  may not be IIDRV, e.g. if  $R_1, \dots, R_n$  are not exponentially distributed and  $e_i$  is reconnected to  $e_0$  more than once while  $e_n$  remains disconnected from  $e_0$  (in case of  $\pi_1$  this may only happen for  $r \geq 2$ ). In consequence, defining  $D(i, n, r, s)$  as the average time during which  $e_i$  remains disconnected from  $e_0$ , one must remember that the successive periods of disconnection may not have one distribution function, therefore in this context "average" cannot be mistaken for "expected value of". The proper meaning of thus defined  $D(i, n, r, s)$  is given by the following lemma.

## Lemma 2

Let  $J_k^{(i)}$  be the number of reconnections between  $e_i$  and  $e_0$  during the interval  $(\varphi_k^{(n)}, \rho_k^{(n)})$ , and let  $\theta_k^{(i)}$  be the total time in  $[\varphi_k^{(n)}, \rho_k^{(n)})$  during which  $e_i$  remains disconnected from  $e_0$ ,  $k \geq 1$ . If the assumptions of Lemma 1 are fulfilled, and  $0 < r_{\min} \leq R_i \leq r_{\max} < \infty$  for  $i = 1, \dots, n$ , then

$$(40) \quad \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{j=1}^m \psi_j^{(i)} \rightarrow_{\text{prob}} \frac{E(\theta_1^{(i)})}{E(J_1^{(i)})}$$

where  $\rightarrow_{\text{prob}}$  denotes convergence in probability.

### Proof

Obviously,  $J_k^{(i)}$  and  $\theta_k^{(i)}$  are IIDRV for  $k \geq 1$ . By virtue of (38),

$$(41) \quad E(\psi_1^{(n)}) \leq E(R_1 + \dots + R_n) \leq n \cdot r_{\max}$$

i.e.  $E(\psi_1^{(n)})$  is finite, thus we have:

$$(42) \quad \begin{aligned} E(\psi_1^{(n)}) &= \int_0^{\infty} [1 - H(t)] dt = \sum_{j=1}^{\infty} \int_{(j-1)r_{\min}}^{jr_{\max}} [1 - H(t)] dt \geq \\ &\geq r_{\min} \sum_{j=1}^{\infty} [1 - H(j \cdot r_{\min})] = r_{\min} \sum_{j=1}^{\infty} \Pr(\psi_1^{(n)} \geq j \cdot r_{\min}) \end{aligned}$$

where  $H$  is the distribution function of  $\psi_1^{(n)}$ . It is also true that:

$$(43) \quad E(J_1^{(i)}) = \sum_{j=1}^{\infty} j \cdot \Pr(J_1^{(i)} = j) = \sum_{j=1}^{\infty} \Pr(J_1^{(i)} \geq j) \leq \sum_{j=1}^{\infty} \Pr(\psi_1^{(n)} \geq j \cdot r_{\min})$$

where the last inequality is a consequence of the following implication: if  $J_1^{(i)} \geq j$ , then at least  $j$  repairs are performed from  $\phi_1^{(n)}$  to  $\rho_1^{(n)}$ . From (42) and (43) we obtain:

$$(44) \quad E(J_1^{(i)}) \leq \frac{n \cdot r_{\max}}{r_{\min}}$$

i.e.  $E(J_1^{(i)})$  is finite.  $E(\theta_1^{(i)})$  is also finite, because  $\theta_1^{(i)} \leq \psi_1^{(n)}$ .

Let  $K(i,m)$ ,  $m \geq 0$ , be an integer valued random variable equal to  $k$  if the interval  $(\varphi_m^{(i)}, \rho_m^{(i)})$  is included in the interval  $(\varphi_k^{(n)}, \rho_k^{(n)})$ , i.e.  $K(i,m)=k$  if  $J_1^{(i)} + \dots + J_{k-1}^{(i)} < m \leq J_1^{(i)} + \dots + J_k^{(i)}$ , where  $J_0^{(i)}=0$ . In consequence of (41)

$$(45) \quad \lim_{m \rightarrow \infty} K(i,m) = \infty$$

holds. For  $m$  such that  $J_1^{(i)} + \dots + J_{K(i,m)-1}^{(i)} > 0$  we have:

$$(46) \quad \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)-1}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)}^{(i)}} \leq \frac{1}{m} \sum_{j=1}^m \psi_j^{(i)} \leq \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)-1}^{(i)}}$$

$$(47) \quad \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)-1}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)}^{(i)}} = \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)}^{(i)}} - \frac{\theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)}^{(i)}}$$

$$(48) \quad \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)-1}^{(i)}} = \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)-1}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)-1}^{(i)}} + \frac{\theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)-1}^{(i)}}$$

From (45) – (48) it follows that:

$$(49) \quad \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{j=1}^m \psi_j^{(i)} = \lim_{m \rightarrow \infty} \frac{\theta_1^{(i)} + \dots + \theta_{K(i,m)}^{(i)}}{J_1^{(i)} + \dots + J_{K(i,m)}^{(i)}} = \lim_{k \rightarrow \infty} \frac{\theta_1^{(i)} + \dots + \theta_k^{(i)}}{k} \cdot \frac{k}{J_1^{(i)} + \dots + J_k^{(i)}}$$

Now (40) is obtained by applying the Khinchin law of large numbers to (49).

Let us now pass to the details of our estimation method. It is based on simulating one sample path of the process  $X_k(i)$  describing the system's behavior, defined in the sequel. Clearly, the component  $e_i$  can be disconnected from or reconnected to  $e_0$  only at the times  $T_k, k \geq 1$ , which are the consecutive moments when any component changes its operational state, i.e. either the component fails or its repair is finished. Thus, the estimation method consists of the following tasks:

- 1) generating the sequence  $\{T_k, k \geq 1\}$ , and the states of all components at the instants  $T_k$ ,
- 2) selecting from  $\{T_k, k \geq 1\}$  these moments when  $e_i$  is disconnected from or reconnected to  $e_0$ , i.e. the moments  $\varphi_j^{(i)}$  and  $\rho_j^{(i)}, j \geq 1$ ,
- 3) estimating  $C(i,n,r,s)$  and  $D(i,n,r,s)$  as the sample means computed from  $\chi_j^{(i)}$  and  $\psi_j^{(i)}, 1 \leq j \leq L$ , where  $L$  is the number of samples.

Task 1 is implemented by Procedure 1, outlined below.

#### Variables used by Procedure 1:

$X_k^{(i)}$ : the state of  $e_i$  at  $T_k$ ; it is assumed that:

$X_k^{(i)} = -q$  if  $e_i$  is the  $q$ -th component in the queue of components awaiting repair,

$X_k^{(i)} = 0$  if  $e_i$  is under repair,

$X_k^{(i)} = 1$  if  $e_i$  is operable and connected to  $e_0$ ,

$X_k^{(i)} = 2$  if  $e_i$  is operable and disconnected from  $e_0$ ,

$i^*$ : index of the failed component, located nearest  $e_0$ ,

$S_k^{(i)}$ : the sojourn time of  $e_i$  in the state  $X_k^{(i)}$ , counted from  $T_k$ , with the assumption that all other components do not change their states before  $e_i$  does,

$q\_len$ : the number of components awaiting repair (queue length),

$avl\_rt$ : the number of available repair teams,

$sim(1,i), sim(0,i)$ : the functions simulating time-to-failure and time-to-repair for  $e_i$ ; the simulation is of Monte Carlo type, therefore it is based on random numbers generation.

## Procedure 1

```
T0 = 0; avl_rt = r; q_len = 0; i* = n + 1;
repeat for i = 1, ..., n {
    X0(i) = 1; S0(i) = sim(1, i);
}

repeat for k ≥ 1
{

    Tk = min(Sk-1(i) : 1 ≤ i ≤ n, Xk-1(i) należy do {0, 1});
    ## If at Tk-1 a component was failed and awaiting repair
    ## or it was operable and disconnected from e0
    ## then the component is irrelevant in determining Tk

    ## adding failed components to the queue (repair policy π1)
    repeat for i = i* - 1, ..., 1
        if (Xk-1(i) = 1 AND Sk-1(i) = Tk) then {
            Xk(i) = -q_len - 1; q_len = q_len + 1; }

    ## releasing repair teams
    repeat for i = 1, ..., n
        if (Xk-1(i) = 0 AND Sk-1(i) = Tk) then {
            Xk(i) = 1; avl_rt = avl_rt + 1; }
```

```

## taking at most avl_rt components for repair
x = avl_rt;
repeat for i = 1,..., n {
  if  $(-x \leq X_k^{(i)} < 0)$  then {
     $X_k^{(i)} = 0$ ; avl_rt = avl_rt - 1; q_len = q_len - 1; }
  if  $(X_k^{(i)} < -x)$  then  $X_k^{(i)} = X_k^{(i)} + x$ ;
}

```

##updating i\* and the states of operable components

```

i* = n + 1;
repeat for i = 1,...,n {
  if  $(X_k^{(i)} \leq 0)$  then i* = i; break; }
repeat for i = 1,..., n {
  if  $(X_k^{(i)} = 1 \text{ AND } i > i^*)$  then  $X_k^{(i)} = 2$ ;
  if  $(X_k^{(i)} = 2 \text{ AND } i < i^*)$  then  $X_k^{(i)} = 1$ ; }

```

##simulating the residual sojourn times of components in their states after  $T_k$

```

repeat for i = 1,..., n {
  if  $(X_k^{(i)} < 0 \text{ OR } X_k^{(i)} = 2)$  then continue;
  if  $(X_k^{(i)} = 0 \text{ AND } X_{k-1}^{(i)} \neq 0)$  then  $S_k^{(i)} = \text{sim}(0,i)$ ;
  if  $(X_k^{(i)} = 0 \text{ AND } X_{k-1}^{(i)} = 0)$  then  $S_k^{(i)} = S_{k-1}^{(i)} - [T_k - T_{k-1}]$ ;
  if  $(X_k^{(i)} = 1 \text{ AND } X_{k-1}^{(i)} \neq 1)$  then  $S_k^{(i)} = \text{sim}(1,i)$ ;
  if  $(X_k^{(i)} = 1 \text{ AND } X_{k-1}^{(i)} = 1)$  then  $S_k^{(i)} = S_{k-1}^{(i)} - [T_k - T_{k-1}]$ ; }

```

} ## end of "repeat for  $k \geq 1$ "

**Remarks:**

1. The time  $S_k^{(i)}$  is simulated only if  $e_i$  changes its state to 0 or 1 at the instant  $T_k$ . Obviously, if  $e_i$  remains in the state 0 or 1, the residual sojourn times for  $e_i$  at  $T_{k-1}$  and  $T_k$  differ by the length of time elapsed from  $T_{k-1}$  to  $T_k$ . If  $e_i$  changes its state to 2 or a negative value, or remains in one of those states, it is irrelevant in determining  $T_{k+1}$ , as neither failure nor repair completion is possible for a component whose state is not 0 or 1.

2. In case of the repair policy  $\pi_2$  the newly failed components are placed before those awaiting repair, hence the following code fragment is used to add failed components to the queue:

```
x = 0;
repeat for i=1,...,j*-1
  if ( $X_{k-1}^{(i)} = 1$  AND  $S_{k-1}^{(i)} = T_k$ ) then {
    x = x+1;  $X_k^{(i)} = -x$ ; q_len = q_len + 1; }
repeat for i=i*,...,n
  if ( $X_{k-1}^{(i)} < 0$ ) then  $X_k^{(i)} = X_{k-1}^{(i)} - x$ ;
```

As follows from the specification of Task 3,  $C(i,n,r,s)$  and  $D(i,n,r,s)$  will be estimated by taking sample means from  $\chi_j^{(i)}$  and  $\psi_j^{(i)}$  over  $L$  operating cycles, where a cycle is the time interval between two consecutive reconnections of  $e_i$  to  $e_0$ , i.e. one of the intervals  $[\rho_{j-1}^{(i)}, \rho_j^{(i)}]$ ,  $j \geq 1$ . Thus, the estimation procedure is constructed by embedding Tasks 2 and 3 into Procedure 1, yielding Procedure 2 outlined below.

### Variables used by Procedure 2:

$j$ : the number of the current cycle,

$T1$  and  $T0$ : the sample values of  $\chi_j^{(i)}$  and  $\psi_j^{(i)}$

$E1$  and  $E0$ : the sample means computed from  $\chi_h^{(i)}$  and  $\psi_h^{(i)}$  over  $h$  varying from 1 to  $j$

$Y_k^{(i)}$ : the state of connection between  $e_i$  and  $e_0$  at  $T_k$ ,  $Y_k^{(i)} = 1$  if  $e_i$  is connected, otherwise  $Y_k^{(i)} = 0$ .

### Procedure 2

$T1 = 0$ ;  $T0 = 0$ ;  $E1 = 0$ ;  $E0 = 0$ ;

$j = 1$ ;

repeat for  $k \geq 1$

{

obtain  $T_k$  and  $X_k^{(1)}, \dots, X_k^{(n)}$  using Procedure 1;

compute  $Y_k^{(i)}$  from  $X_k^{(1)}, \dots, X_k^{(n)}$

if  $(Y_{k-1}^{(i)} \text{ EQ } 1)$  then {

$T1 = T1 + (T_k - T_{k-1})$ ;

if  $(Y_k^{(i)} \text{ EQ } 0)$  ##  $e_i$  is disconnected from  $e_0$  at  $T_k$

then {

$E1 = E1 \cdot (j - 1) / j + T1 / j$ ; ## updating  $E1$  during the cycle  $j$

$T1 = 0$ ; }

}



```

if ( $Y_{k-1}^{(i)}$  EQ 0) {
    T0 = T0 + ( $T_k - T_{k-1}$ );
    if ( $Y_k^{(i)}$  EQ 1) ##  $e_i$  is reconnected to  $e_0$  at  $T_k$ 
    then {
        E0 =  $E0 \cdot (j - 1) / j + T0 / j$ ; ## updating E0 at the end of the cycle j
        T0 = 0;
        j = j + 1;
        if (j GT L) then terminate; }
    }

```

} ## end of "repeat for  $k \geq 1$ "

### Remarks:

1. E1 and E0 are updated based on the following formula:

$$(50) \quad \mu_{n+1} = \mu_n \cdot n / (n+1) + x_{n+1} / (n+1)$$

where

$$(51) \quad \mu_n = (x_1 + \dots + x_n) / n$$

2. In the step k only the values  $X_{k-1}^{(i)}$ ,  $X_k^{(i)}$ ,  $T_{k-1}$ ,  $T_k$  are used to update T0, T1, and possibly E0, E1, while the analogous values obtained in the steps 1, ..., k-2 are irrelevant. In consequence, it is necessary to store only the X's and T's obtained in the current and the previous step of Procedure 2.

## 5. Some numerical results

In Tables 1 and 3 several results obtained using Procedure 2 are presented. It is assumed that  $L_i$  and  $R_i$  are exponentially distributed with  $\lambda_i = 0.01$ ,  $\mu_i = 0.1$ ,  $1 \leq i \leq n$ . The time unit is one hour. The results presented in Table 2 are obtained from (16), (22), (23), and (37).

Tab. 1. Estimated average disconnection periods for a two-component system

$D(i,2,r,s)$	$r=1, s=1$	$r=1, s=2$	$r=2, s=1$	$r=2, s=2$
$i=1$	10.84	10.84	10.01	10.01
$i=2$	10.46	10.45	10.23	10.24

Tab. 2. Exact average disconnection periods for a two-component system

$D(i,2,r,s)$	$r=1, s=1$	$r=1, s=2$	$r=2, s=1$	$r=2, s=2$
$i=1$	10.8(3)	10.8(3)	10.00	10.00
$i=2$	10.(45)	10.(45)	10.2381	10.2381

Tab. 3. Estimated average disconnection periods for a ten-component system

$D(i,10,r,s)$	$r=1, s=1$	$r=1, s=2$	$r=2, s=1$	$r=2, s=2$
$i=1$	16.60	15.23	10.43	10.35
$i=5$	15.20	15.03	11.14	11.12
$i=10$	13.95	14.11	12.10	12.11

It is interesting to see that if there is one repair team then  $D(i,n,r,s)$  is shorter for  $i=n$  than for  $i=1$ . This observation is confirmed by the exact results for  $n=2$ . To explain this fact note that  $e_n$  can fail only if  $e_1, \dots, e_{n-1}$  are all in operating state. Thus the repair of  $e_n$  begins immediately after its failure, while  $e_1$  has to wait if another component is under repair at the moment of  $e_1$ 's failure.

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the 1990s, the number of people with diabetes has increased in all industrialized countries. In the Netherlands, the prevalence of diabetes is estimated to be 6.5% in 1995, which corresponds to 1.5 million people (1).

Diabetes is a chronic disease with a high prevalence and a high mortality. The most common complications of diabetes are cardiovascular disease, nephropathy, retinopathy, and neuropathy. The prevalence of these complications is high, and the mortality is also high. In the Netherlands, the mortality of diabetes is estimated to be 10% per year (2).

The most common complication of diabetes is cardiovascular disease. The prevalence of cardiovascular disease is high, and the mortality is also high. In the Netherlands, the mortality of cardiovascular disease is estimated to be 10% per year (3). The most common complication of cardiovascular disease is coronary artery disease. The prevalence of coronary artery disease is high, and the mortality is also high. In the Netherlands, the mortality of coronary artery disease is estimated to be 10% per year (4).

The most common complication of coronary artery disease is myocardial infarction. The prevalence of myocardial infarction is high, and the mortality is also high. In the Netherlands, the mortality of myocardial infarction is estimated to be 10% per year (5). The most common complication of myocardial infarction is heart failure. The prevalence of heart failure is high, and the mortality is also high. In the Netherlands, the mortality of heart failure is estimated to be 10% per year (6).

The most common complication of heart failure is stroke. The prevalence of stroke is high, and the mortality is also high. In the Netherlands, the mortality of stroke is estimated to be 10% per year (7). The most common complication of stroke is dementia. The prevalence of dementia is high, and the mortality is also high. In the Netherlands, the mortality of dementia is estimated to be 10% per year (8).

The most common complication of dementia is depression. The prevalence of depression is high, and the mortality is also high. In the Netherlands, the mortality of depression is estimated to be 10% per year (9). The most common complication of depression is suicide. The prevalence of suicide is high, and the mortality is also high. In the Netherlands, the mortality of suicide is estimated to be 10% per year (10).

The most common complication of suicide is death. The prevalence of death is high, and the mortality is also high. In the Netherlands, the mortality of death is estimated to be 10% per year (11). The most common complication of death is burial. The prevalence of burial is high, and the mortality is also high. In the Netherlands, the mortality of burial is estimated to be 10% per year (12).

The most common complication of burial is cremation. The prevalence of cremation is high, and the mortality is also high. In the Netherlands, the mortality of cremation is estimated to be 10% per year (13). The most common complication of cremation is ash. The prevalence of ash is high, and the mortality is also high. In the Netherlands, the mortality of ash is estimated to be 10% per year (14).

the 1990s, the number of people in the UK who are aged 65 and over has increased from 10.5 million to 13.5 million (15.5% of the population).

There is a growing awareness of the need to address the needs of older people, and the Government has set out a strategy for the 21st century in the White Paper on *Ageing Better: The Government's Strategy for Older People* (Department of Health 1999). This strategy is based on the following principles:

- Older people should be able to live independently and actively in their own homes.
- Older people should be able to live in their own communities.
- Older people should be able to live in their own homes and communities for as long as possible.

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