
SHOCK MODELS UNDER POLICY N

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ABSTRACT

We present the life distribution of a device subject to shocks governed by phase-type distributions. The probability of failures after shocks follows discrete phase-type distribution. Lifetimes between shocks are affected by the number of cumulated shocks and they follow continuous phase-type distributions. The device can support a maximum of N shocks. We calculate the distribution of the lifetime of the device and illustrate the calculations by means of a numerical application. Computational aspects are introduced. This model extends other previously considered in the literature.

1 INTRODUCTION AND BACKGROUND

The classical shocks model of Esary et al. [2] study the lifetime of a device subject to shocks that arrive randomly following a Poisson process $\{N(t), t \geq 0\}$. The device has a probability \bar{P}_k of survival to k shocks. Then, the survival function of the model, $H(t)$, is given by:

$$\bar{H}(t) = \sum_{k=0}^{\infty} P\{N(t) = k\} \bar{P}_k \quad (1)$$

This model has been studied under non-parametric methodology by different authors, considering reliability classes. Neuts et al. (1981), introduced phase-type distributions and calculated explicitly the lifetime distribution of the device. Manoharan et al. (1992) considered a finite mixture of homogeneous Poisson process as arrival process. In these previous papers, the number of shocks that arrives to the device is unlimited. In Neuts et al. (2000) phase-distributions are used to study a model subject to a limited number of failures.

We present a model limiting the number of shocks that the device can stand. The probability of failure due to the shocks follow a discrete phase-type distribution. The interarrival times between shocks depend on the number of cumulated shocks.

The process that governs the system is a Markov one with vectorial state space. We calculate the lifetime distribution of the device and present a numerical example illustrating the calculations.

In Section 2 the shock model is presented. In Section 3 the Markov model that governs the system is constructed, and the lifetime distribution of the device determined. In Section 4 a numerical application is performed.

Given that the phase-type distributions play a fundamental role throughout the paper, we define them in the discrete and continuous cases. More details about these distributions can be seen in Neuts (1981).

2 DEFINITIONS

Definition 1. The distribution $H(\cdot)$ on $[0, \infty[$ is a phase-type distribution (PH-distribution) with representation (α, T) if it is the distribution of the time until the absorption in a Markov process on the states $\{1, \dots, m, m+1\}$ generator

$$\begin{pmatrix} T & T^0 \\ 0 & 0 \end{pmatrix}$$

and initial row probability vector α of order m . We assume that the states $\{1, \dots, m\}$ are all transient. Throughout this paper e denotes a column vector with all components equal to one the dimension of which is determined by the context. The matrix T of order m is non-singular with negative diagonal entries and non-negative off-diagonal entries and satisfies $-Te = T^0 \geq 0$. The distribution of $H(\cdot)$ is given by,

$$H(x) = 1 - \alpha \exp(Tx)e, \quad x \geq 0$$

It will be denoted that $H(\cdot)$ follows $\text{PH}(\alpha, T)$ distribution.

Definition 2. A density $\{p_k\}$ of the nonnegative integers is of phase type if and only if there exists a finite Markov chain with transition probability matrix P of order $n+1$ of the form

$$\begin{pmatrix} S & S^0 \\ 0 & 1 \end{pmatrix}$$

and initial probability vector (β, β_{n+1}) , where β is a row n -vector. Here S is a substochastic matrix such that $Se + S^0 = e$, and $(I-S)$ is non-singular. The density of the time until absorption is given by

$$\begin{aligned} p_0 &= \beta_{n+1}, \\ p_k &= \beta S^{k-1} S^0, \quad \text{for } k \geq 1 \end{aligned}$$

It will be denoted that $\{p_k\}$ follows a $\text{PH}_d(\beta, S)$.

We use the Kronecker product (see [1]).

3 SHOCK MODEL

Suppose that a device is subjected to shocks according to the following assumptions.

1. Let $X^{(k)}$ be the interarrival times between the shocks k th and $(k+1)$ th, $k=0, 1, \dots$. These random times follow distributions $\text{PH}(\beta^{(k)}, S^{(k)})$ of order n_k .
2. The device can accumulate a maximum of N shocks, in such a way that it is replaced by a new one to the arrival of the $N+1$ shock. We denote by p_k the probability of failure of the device due to the arrival of the k th shock, $k \geq 1$. We assume that p_k follows a distribution $\text{PH}_d(\gamma, L)$ of order $N+1$.

This representation is given by $\gamma = (1, 0, \dots, 0)$ and

$$L = \begin{pmatrix} 0 & l_1 & 0 & \dots & 0 \\ 0 & 0 & l_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & l_N \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix}; L^0 = \begin{pmatrix} 1-l_1 \\ 1-l_2 \\ \vdots \\ 1-l_N \\ 1 \end{pmatrix}$$

The entries l_k , $k=1,\dots,N$, denote the conditional probability that the system will survive to the k th shock, given that it has survived to the $(k-1)$ th shock. These l_k are useful to find a representation to the distribution $\{p_k\}$. It is clear that

$$p_k = \gamma L^{k-1} L^0, k \geq 1$$

The survival probability to the k th shock is

$$\bar{P}_k = \sum_{v=k}^{N+1} p_v = \gamma (L^k - L^{N+1}) e, k=1,\dots,N; \bar{P}_0 = 1 \quad (2)$$

Under the assumptions the survival function (1) follows a PH-distribution, that will be calculated.

4 MARKOV MODEL

Under these assumptions, the probabilistic model that governs the system is a Markov process. The occupied exponential states by the device will be denoted by (k,i) , k being the number of cumulated shocks, i the phase of the random variable $X^{(k)}$. We group these states in sets, named macro-states, that will be denoted by k , $k=0,1,\dots,N$. The number of exponential states of the macro-state k is n_{k+1} . The infinitesimal generator of the Markov process is built in terms of the transition between macro-states, and, consequently, it will be a generator formed by blocks.

We denote by T_k the lifetime of the device when the failure occurs at the arrival of the $(k+1)$ th shock. It is clear that

$$T_k = \sum_{i=0}^k X^{(i)}, 0 \leq k \leq N \quad (3)$$

This random variable is the sum of independent random variables PH-distributed and follows a distribution $\text{PH}(g^{(k)}, G^{(k)})$. We calculate this representation. If the device fails at the $(k+1)$ th shock, it has survived to the first k shocks. Thus, the transitions between the up macro-states to the occurrence of the failure are $0 \rightarrow 1 \rightarrow 2 \rightarrow \dots \rightarrow k$. These transitions $j \rightarrow j+1$, $j=0,1,\dots,k$, occur when a non-fatal shock arrives being the device in the macro-state j , and these are governed by the absorption vector $S^{(j)}$. Then, the new interarrival period initiates following the initial vector $\beta^{(j)}$. Thus, the matrix $G^{(k)}$ is

$$G^{(k)} = \begin{pmatrix} S^{(0)} & S^{0(0)}\beta^{(1)} & & & \\ & S^{(1)} & S^{0(1)}\beta^{(2)} & & \\ & & \ddots & & \\ & & & & S^{(k)} \end{pmatrix}$$

We assume that the device initiates with 0 shocks, so the initial vector $g^{(k)}$ is given by

$$g^{(k)} = (\beta^{(0)}, 0, \dots, 0), k = 0, 1, \dots, N$$

Denoting by T the lifetime of the system, we have

$$R(t) = P(T > t) = \sum_{k=0}^N p_{k+1} P(T_k > t), t \geq 0 \tag{4}$$

We determine the distribution of the random variable T as follows. This is the distribution of a finite mixture of PH-distributions, it is a PH-distribution with well-known representation (see [4]) given by (v, V) , with,

$$v = (p_1 g^{(0)}, p_2 g^{(1)}, \dots, p_{N+1} g^{(N)}), V = \begin{pmatrix} G^{(0)} & & & \\ & G^{(1)} & & \\ & & \ddots & \\ & & & G^{(N)} \end{pmatrix} \tag{5}$$

The analytic expression of the survival function of T is:

$$P(T > t) = v \exp \{Vt\} e = \sum_{k=0}^N p_{k+1} g^{(k)} \exp(G^{(k)}t) e, t \geq 0 \tag{6}$$

5 NUMERICAL APPLICATION

We assume a system that can stand a maximum of 4 shocks, then the system is replaced to the arrival of the 4th shock. The probabilities of failure to the arrival of shocks are $p_0=0, p_1=0.2, p_2=0.2, p_3=0.3, p_4=0.3$. The shocks arrival time follow Erlang distributions with the following PH representations:

– $X^{(0)} \rightsquigarrow PH(\beta^{(0)}, S^{(0)})$, with

$$\beta^{(0)} = (1, 0), S^{(0)} = \begin{pmatrix} -1 & 1 \\ & -1 \end{pmatrix}, \text{ sum of two identical exponentials with parameter } \lambda=1.$$

– $X^{(1)} \rightsquigarrow PH(\beta^{(1)}, S^{(1)})$, with

$\beta^{(0)} = (1, 0), S^{(0)} = \begin{pmatrix} -3 & 3 \\ & -3 \end{pmatrix}$, sum of two identical exponentials with parameter $\lambda=3$.

- $X^{(2)} \sim PH(\beta^{(2)}, S^{(2)})$, with

$\beta^{(0)} = (1, 0, 0), S^{(0)} = \begin{pmatrix} -1 & 1 & \\ & -1 & 1 \\ & & -1 \end{pmatrix}$, sum of three identical exponentials with parameter $\lambda=1$.

- $X^{(3)} \sim PH(\beta^{(3)}, S^{(3)})$, with

$\beta^{(0)} = (1, 0, 0), S^{(0)} = \begin{pmatrix} -2 & 2 & \\ & -2 & 2 \\ & & -2 \end{pmatrix}$, sum of three identical exponentials with parameter $\lambda=2$.

The values of these distributions are simulated following these steps. We have performed a total of one hundred simulated values of the exponential distributions corresponding to the different parameters above assigned. Adding these values we calculated simulated values of the random variables X 's, and using these values we calculated simulated values of the device lifetime to k th shock $T_k, k=1,2,3$. We do not define the unit time.

- $T^{(0)} \sim PH(g^{(0)}, G^{(0)})$, with $g^{(0)} = \beta^{(0)}, G^{(0)} = S^{(0)}$

- $T^{(1)} \sim PH(g^{(1)}, G^{(1)})$, with $g^{(1)} = (\beta^{(0)}, 0, 0), G^{(1)} = \begin{pmatrix} -1 & 1 & \\ & -1 & 1 \\ & & -3 & 3 \\ & & & -3 \end{pmatrix}$

- $T^{(2)} \sim PH(g^{(2)}, G^{(2)})$, with $g^{(2)} = (\beta^{(0)}, 0, 0, 0, 0, 0), G^{(1)} = \begin{pmatrix} -1 & 1 & & & & \\ & -1 & 1 & & & \\ & & -3 & 3 & & \\ & & & -3 & 3 & \\ & & & & -1 & 1 \\ & & & & & -1 & 1 \\ & & & & & & -1 \end{pmatrix}$

- $T^{(3)} \sim PH(g^{(3)}, G^{(3)})$, with $g^{(3)} = (\beta^{(0)}, 0, 0, 0, 0, 0, 0, 0, 0)$, and,

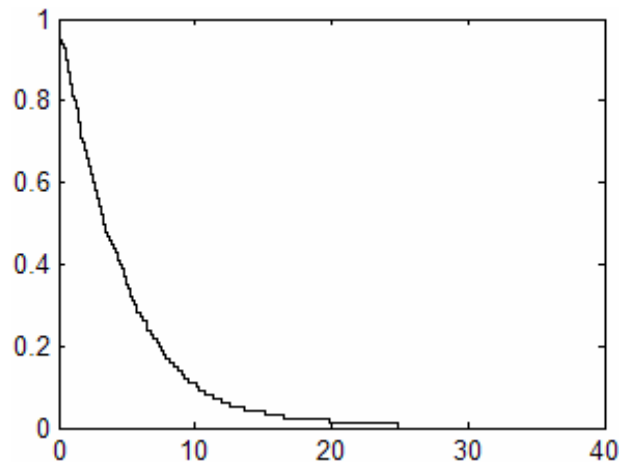


Figure 2. Simulation of the Survival function of the system.

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