## A Method for Evaluating the Distribution of the Total Cost of a Random Process over its Lifetime

by

Phil Pollett (UQ) Valerie Stefanov (UWA)



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  - A set of states A
  - The time  $\tau$  to first exit from A
  - The cost (per unit time)  $f_x$  of being in state x
  - The "path integral"

$$\Gamma = \int_0^\tau f_{X(t)} \, dt,$$

the total cost incurred before leaving A

## Examples

Let X(t) be the water level in a dam at time t. If τ is the time the dam first empties, and f<sub>x</sub> = 1<sub>{x>l}</sub>, then Γ is the total time that the level is above l:

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• Let (I(t), S(t)) be the number of infectives and susceptibles in an epidemic at time t. If  $\tau$  is the period of infection and  $f_{(i,s)} = i$ , then  $\Gamma$  is the total amount of infection:

$$\Gamma = \int_0^\tau I(t) \, dt.$$

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Our problem is to determine the *expected total cost*, and the *distribution of the total cost*.

For simplicity, suppose that X(t) takes values in  $S = \{0, 1, ...\}$ . For example, X(t) might be the number in a population at time t, and  $A = \{1, 2, ...\}$ , so the  $\tau$  is the time to extinction.



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 and  $T_j = \sum_{n=1}^{N_j} X_{jn}$ ,

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where  $X_{jn}$ , n = 1, 2, ..., are the successive occupancy times for state j. If these times are independent and identically distributed, then  $E(\Gamma) = \sum_{j \in A} f_j E(N_j) \mu_j$ , where  $\mu_j$  is the mean occupancy time for state j.

## Markovian models

• We will assume that  $(X(t), t \ge 0)$  is a *Markov chain* with *transition rates*  $Q = (q_{ij}, i, j \in S)$ , so that  $q_{ij}$  represents the rate of transition from state i to state j, for  $j \ne i$ , and  $q_{ii} = -q_i$ , where  $q_i := \sum_{j \ne i} q_{ij}$  (<  $\infty$ ) represents the total rate out of state i.

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$$Q = \begin{pmatrix} -\lambda_0 & \lambda_0 & 0 & 0 & \cdots \\ \mu_1 & -(\lambda_1 + \mu_1) & \lambda_1 & 0 & \cdots \\ 0 & \mu_2 & -(\lambda_2 + \mu_2) & \lambda_2 & \cdots \\ \vdots & \vdots & \vdots & 0 & \ddots \end{pmatrix}$$

#### The expected value of $\Gamma$

• Let  $e_i = E_i(\Gamma) := E(\Gamma|X(0) = i)$ , and condition on the time of the first jump and the state visited at that time, to get

$$E_{i}(\Gamma) = \int_{0}^{\infty} \sum_{k \neq i} \left( \frac{f_{i}}{q_{i}} + E_{k}(\Gamma) \right) \frac{q_{ik}}{q_{i}} q_{i} e^{-q_{i}u} du,$$

which leads to

$$q_i e_i = f_i + \sum_{k \neq i} q_{ik} e_k,$$

so that

$$\sum_{k} q_{ik}e_k + f_i = 0. \qquad (Q\boldsymbol{e} = -\boldsymbol{f})$$

We can do better:

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**Theorem 1**  $e = (e_i, i \in A)$ , where  $e_i = E_i(\Gamma)$ , is the *minimal* non-negative solution to

$$\sum_{k \in A} q_{ik} z_k + f_i = 0, \quad i \in A, \qquad (Q \boldsymbol{z} + \boldsymbol{f} = \boldsymbol{0})$$

in the sense that *e* satisfies these equations, and, if  $z = (z_i, i \in A)$  is any non-negative solution, then  $e_i \leq z_i$  for all  $i \in A$ .

## **Birth-death processes**

Assume that the birth rates (λ<sub>i</sub>, i ≥ 1) and the death rates (μ<sub>i</sub>, i ≥ 0) are all strictly positive, except that λ<sub>0</sub> = 0. So, all states in A = {1, 2, ...} intercommunicate, and 0 is an absorbing state (corresponding to population extinction).

Define *potential coefficients*  $(\pi_i, i \ge 1)$  by  $\pi_1 = 1$  and

$$\pi_i = \prod_{j=2}^i \frac{\lambda_{j-1}}{\mu_j}, \qquad i \ge 2,$$

and assume that  $A := \sum_{i=1}^{\infty} 1/(\mu_i \pi_i)$  diverges, a condition that corresponds to extinction being certain.

On applying Theorem 1 we get:

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**Proposition** The expected cost up to the time of extinction is given by

$$E_i(\Gamma) = \sum_{j=1}^i \frac{1}{\mu_j \pi_j} \sum_{k=j}^\infty f_k \pi_k,$$

for all  $i \ge 1$ , this being finite if and only if  $\sum_{k=1}^{\infty} f_k \pi_k < \infty$ .

## The distribution of $\Gamma$

• Let  $y_i(\theta) = E_i(e^{-\theta\Gamma})$  be the Laplace-Steiltjes transform of the distribution of  $\Gamma$ :

$$y_i(\theta) = \int_0^\infty e^{-\theta x} d\Pr(\Gamma \le x | X(0) = i).$$

A similar argument leads to:

**Theorem 2** For each  $\theta > 0$ ,  $y(\theta) = (y_i(\theta), i \in S)$  is the *maximal* solution to

$$\sum_{k \in S} q_{ik} z_k = \theta f_i z_i, \quad i \in A,$$

with  $0 \le z_i \le 1$  for  $i \in A$  and  $z_i = 1$  for  $i \notin A$ .

Assume that the transition rates have the form

$$q_{ij} = \begin{cases} i\rho a, & i \ge 0, \ j = i+1, \\ -i\rho, & i \ge 0, \ j = i, \\ i\rho d_{i-j}, & i \ge 2, \ 1 \le j < i, \\ i\rho \sum_{k\ge i} d_k, & i \ge 1, \ j = 0, \end{cases}$$

with all other transition rates equal to 0. Here  $\rho$  and a are positive,  $d_i$  is positive for at least one i in  $A = \{1, 2, ...\}$ and  $a + \sum_{i=1}^{\infty} d_i = 1$ .

Clearly 0 is an absorbing state for the process and A is a communicating class.

• We will consider only the *subcritical case*, where the drift D, given by  $D = a - \sum_{i=1}^{\infty} id_i$ , is strictly negative and extinction is certain. Let b(s) = d(s) - s, where d is the probability generating function  $d(s) = a + \sum_{i=1}^{\infty} d_i s^{i+1}$ , |s| < 1. We can evaluate  $E_i(e^{-\theta\Gamma})$  for specific choices of f. Take  $f_i = i$ . We seek the maximal solution to

$$\sum_{j=0}^{\infty} q_{ij} z_j = \theta i z_i, \qquad i \ge 1,$$

satisfying  $0 \le z_i \le 1$  for  $i \ge 1$  and  $z_0 = 1$ .

We will consider only the *subcritical case*, where the drift *D*, given by *D* = *a* − ∑<sub>i=1</sub><sup>∞</sup> *id<sub>i</sub>*, is strictly negative and extinction is certain. Let *b*(*s*) = *d*(*s*) − *s*, where *d* is the probability generating function *d*(*s*) = *a* + ∑<sub>i=1</sub><sup>∞</sup> *d<sub>i</sub>s<sup>i+1</sup>*, |*s*| < 1. We can evaluate *E<sub>i</sub>*(*e*<sup>−θΓ</sup>) for specific choices of *f*. Take *f<sub>i</sub>* = *i*. We seek the maximal solution to

$$\rho a z_{i+1} - \rho z_i + \rho \sum_{j=1}^{i-1} d_{i-j} z_j + \rho z_0 \sum_{j=i}^{\infty} d_j = \theta z_i, \quad i \ge 1,$$

satisfying  $0 \le z_i \le 1$  for  $i \ge 1$  and  $z_0 = 1$ .

• Multiplying by  $s^{i-1}$  and summing over i gives

$$\sum_{i=1}^{\infty} E_i(e^{-\theta\Gamma})s^{i-1} = \frac{1}{1-s} - \frac{\theta(\gamma_{\theta} - s)}{(1-\gamma_{\theta})(1-s)(\rho b(s) - \theta s)},$$

where  $\gamma_{\theta}$  is the unique solution to  $\rho b(s) = \theta s$  on the interval  $0 < s < \sigma$ . In the case of "geometric catastrophes" ( $d_i = d(1-q)q^{i-1}$ ,  $i \ge 1$ , where d > 0 satisfies a + d = 1, and  $0 \le q < 1$ ), we get

$$E_i(e^{-\theta\Gamma}) = \frac{\beta(\theta) - q}{1 - q} \left(\beta(\theta)\right)^{i-1}, \quad i \ge 1,$$

where  $\beta(\theta)$  is the smaller of the two zeros of  $a\rho s^2 - (\rho(1+qa)+\theta)s + \rho(d+qa) + q\theta$ .