

**MATH-449 - Biostatistics**  
**EPFL, Spring 2025**  
**Problem Set 11**

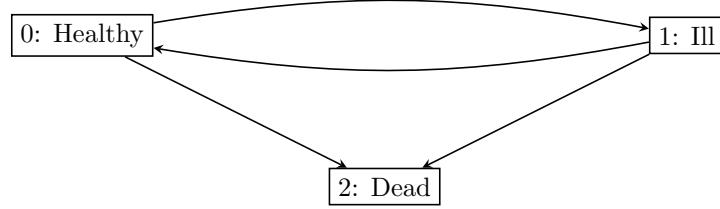


Figure 1: Illness-death model

1. A common example of a multi-state process is an illness-death process. The transition diagram in Figure 1 shows an illness-death process  $\{X(t)\}_t$  where people are either healthy (state 0), ill (state 1) or dead (state 2) at any given time  $t$ . The arrows indicate the possible transitions: healthy people may become ill and ill people may recover to become healthy as time progresses. All individuals are at risk of dying, i.e., transferring into state 2. Thus,  $X(t) \in \{0, 1, 2\}$  for all  $t$ .
  - (a) We will first consider the time until the *first* event of healthy individuals. We let  $\kappa = 2$  signify that the first event is death, and  $\kappa = 1$  signify that the first event is illness. We let  $T^f$  be the first occurrence of either illness or death and consider an i.i.d. sample

$$\{(\tilde{T}_i^f, D_i^f, \kappa_i)\}_{i=1}^n$$

of healthy individuals where

$$\begin{aligned} D_i^f &= I(T_i^f \leq \tilde{T}_i^f), \\ \tilde{T}_i^f &= \min(T_i^f, T_i^*), \\ N_i^j(t) &= I(\tilde{T}_i^f \leq t, D_i^f = 1, \kappa_i = j), \\ Z_i^f(t) &= I(\tilde{T}_i^f \geq t), \\ \beta^j(t) &= \lim_{h \rightarrow 0^+} \frac{1}{h} P(t \leq T^f < t + h, \kappa = j \mid t \leq T^f). \end{aligned}$$

Let  $N^j(t) = \sum_{i=1}^n N_i^j(t)$  and  $Z^f(t) = \sum_{i=1}^n Z_i^f(t)$ , and suppose that  $T^f \perp T^*$  holds, i.e. that a random censoring assumption is satisfied. Answer yes/no on the following claim, and give an explanation for your answer:

The estimator  $\prod_{T_i^f \leq t} \left\{ 1 - \frac{\Delta N^2(T_i^f)}{Z^f(T_i^f)} \right\}$  is an estimator of the survival function at time  $t$ .

- (b) We will now return to the illness-death process  $X$  shown in the diagram in Figure 1. In the lectures we defined the transition intensities  $\alpha_{ij}(t)$  by

$$\alpha_{ij}(t) = \lim_{h \rightarrow 0^+} \frac{1}{h} P_{ij}(t, t + h) = \lim_{h \rightarrow 0^+} \frac{1}{h} P(X(t + h) = j \mid X(t) = i)$$

for transitioning from state  $i$  to state  $j$  when  $i \neq j$ , and  $\alpha_{ii}(t) = -\sum_{j \neq i} \alpha_{ij}(t)$ . Use the definition of the transition intensity matrix,

$$\boldsymbol{\alpha}(t) = \lim_{h \rightarrow 0^+} \frac{1}{h} (\mathbf{P}(t, t + h) - \mathbf{I}),$$

to show that the  $(i, j)$ 'th element of  $\boldsymbol{\alpha}(t)$  is  $\alpha_{ij}(t)$ , i.e. that  $(\boldsymbol{\alpha}(t))_{ij} = \alpha_{ij}(t)$ .

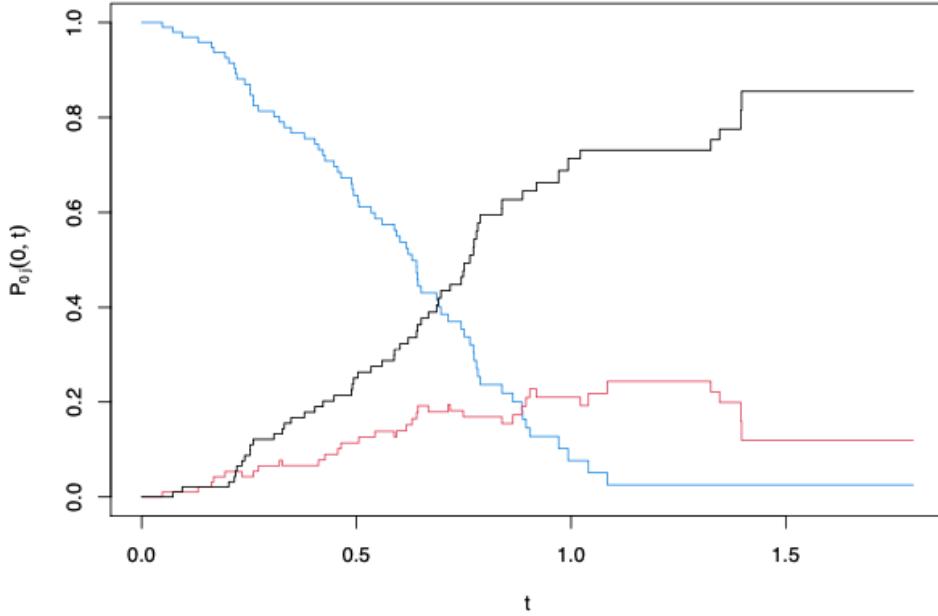


Figure 2: Plot of estimated transition probabilities based on a simulated illness death model.

(c) Is  $\alpha_{ij}(t)$  zero for any  $(i, j)$  in the example shown in the graph in Figure 1? Write out the matrix  $\alpha(t)$ . Use the Kolmogorov forward equation to express  $(P_{00}(s, t), P_{01}(s, t))^T$  as an integral equation on the form

$$\begin{pmatrix} P_{00}(s, t) \\ P_{01}(s, t) \end{pmatrix} = \begin{pmatrix} f^0(s, t) \\ f^1(s, t) \end{pmatrix} + \int_s^t G(t, u) \begin{pmatrix} P_{00}(s, u) \\ P_{01}(s, u) \end{pmatrix} du$$

where  $G$  is a  $2 \times 2$  matrix. What is  $f^0$ ,  $f^1$  and  $G$ ?

(d) Attached is a simulation of an illness-death process where all individuals started in state 0 (healthy). Estimates of the transition probabilities  $P_{0j}(0, t)$  for  $j \in \{0, 1, 2\}$ , obtained from the Aalen-Johansen estimator, are plotted over time  $t$  in Figure 2. State which of the curves (blue, black, red) correspond to which of the quantities  $P_{00}$ ,  $P_{01}$  and  $P_{02}$ . Explain your reasoning.

(e) The transition intensity matrix  $\alpha$  used for generating Figure 2 contains one zero more than the transition intensity matrix you wrote down in point (c). That is, there is a combination  $(i, j)$  such that  $\alpha_{ij} = 0$  in the data generating model that gave rise to Figure 2, but the diagram in Figure 1 suggests that particular  $\alpha_{ij}$  is not zero. State which combination  $(i, j)$  it is, and explain why Figure 2 indicates that the transition intensity  $\alpha_{ij}(t)$  is zero.

2. Compute  $P(Z = 0|X_0 = 10, Y_0 = 1)$ ,  $P(Z = 1|X_0 = 10, Y_0 = 1)$ ,  $P(Z = 2|X_0 = 10, Y_0 = 1)$  and  $P(Z = 3|X_0 = 10, Y_0 = 1)$  for the Reed-Frost model.

**Hint:** Compute the probability for each epidemic chain giving  $Z = k$  separately and then add these probabilities.

3. Show that  $y(t) \rightarrow 0$  as  $t \rightarrow \infty$  for the model of Kermack and McKendrick given by the differential equations

$$\begin{aligned}x'(t) &= -\lambda x(t)y(t) \\y'(t) &= \lambda x(t)y(t) - \gamma y(t) \\z'(t) &= \gamma y(t),\end{aligned}$$

for some  $\lambda, \gamma > 0$