

## Appendix

### Appendix A

Nonparametric estimation of  $H_g(t)$

Define  $N_{ig}(t) = I(X_i \geq t, \delta_i = 1, G = g)$ ,  $i = 1, 2, \dots, n$ ,  $g \in G$ , where  $G$  contains all the non-empty subsets of  $\{1, 2, \dots, k\}$  and  $Y_i(t) = I(X_i \leq t)$ .

For left censored data, we have

$$P(X_i \in (t - dt, t), \delta_i = 1, G = g | F_{t+}) = \begin{cases} h_g(t)dt & \text{if } X_i \leq t \\ = 0 & \text{if } X_i > t \end{cases} \quad (\text{A.1})$$

which leads to the fact that,

$$E[dN_{ig}(t) | F_{t+}] = Y_i(t)h_g(t)dt \quad (\text{A.2})$$

where  $dN_{ig}(t) = I(X_i = t, \delta_i = 1, J_i = j)$ .

Denote  $N_g(t) = \sum_{i=1}^n N_{ig}(t)$ , and  $Y(t) = \sum_{i=1}^n I(X_i \leq t)$ . Now we consider the counting process martingale,

$$M_g(t) = N_g(t) - A_g(t) \quad (\text{A.3})$$

where  $A_g(t) = \sum_{i=1}^n \int_t^{t_0} I(X_i \leq u)h_g(u)du$  and  $t_0 = \inf\{t; F(t) < 1\}$ . We also have

$$E[N_g(t) | F_{t+}] = E[A_g(t) | F_{t+}] = A_g(t) \quad (\text{A.4})$$

and

$$E[dA_g(s) | F_{s+}] = E[-I(X \leq s) | F_{s+}] = dA_g(s). \quad (\text{A.5})$$

From (A.3), (A.4), and (A.5), we have

$$E[dM_g(t) | F_{t+}] = E[dN_g(t) - dA_g(t) | F_{t+}] = 0. \quad (\text{A.6})$$

If  $E[dM_g(t) | F_{t+}] = 0$ , then for all  $t \leq s$

$$\begin{aligned} E[M_g(t) | F_s] - M_g(s) &= E[M_g(t) - M_g(s) | F_s] \\ &= E\left[\int_t^s dM_g(u) | F_s\right] \\ &= \int_t^s E[E dM_g(u) | F_{u+} | F_s] = 0. \end{aligned} \quad (\text{A.7})$$

Thus (A.7) proves that  $M_g(t)$  is a martingale. From (A.3) we can write

$$dN_g(t) = Y(t)h_g(t)dt + dM_g(t). \quad (\text{A.8})$$

If  $Y(t) > 0$ , then we have,

$$\frac{dN_g(t)}{Y(t)} = h_g(t)dt + \frac{dM_g(t)}{Y(t)}. \quad (\text{A.9})$$

If  $dM_g(t)$  is noise, then so is  $\frac{dM_g(t)}{Y(t)}$ , because value of  $Y(t)$  at time  $t$  are known at time  $t+$ . We have  $E\left[\frac{dM_g(t)}{Y(t)} | F_{t+}\right] = 0$ .

Let  $C(t) = I(Y(t) > 0)$ . Integrating both sides of (A.9) we get

$$\int_t^{t_0} \frac{C(u)dN_g(u)}{Y(u)} = \int_t^{t_0} C(u)h_g(u)du + \int_t^{t_0} \frac{C(u)dM_g(u)}{Y(u)} \quad (\text{A.10})$$

Now from (A.9) the estimator of cumulative reversed hazard rate at time  $t$  with  $g$  observed as set of possible causes  $H_g(t)$  is obtained as

$$\hat{H}_g(t) = \int_t^{t_0} \frac{C(u)dN_g(u)}{Y(u)} \quad (\text{A.11})$$