

# Variance of $m()$ , questions in the paper

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We denote  $T_i, i = 1, \dots, N$  are the independent, identically, distributed (iid) lifetimes, whose corresponding cumulative distribution function (CDF) is  $F$ , probability distribution function (PDF) is  $f$ ; the censoring time is defined as  $C_i, i = 1, \dots, N$ .  $C_i$ s are also iid, with CDF denoted as  $G$  and PDF denoted as  $g$ . We set the censors happen on the right and the observed time is  $Z_i = T_i \wedge C_i$ , whose CDF is  $H$  and PDF is  $h$ . The  $\delta_i = I_{[T_i \leq C_i]}$  is the status indicator, which shows whether subject  $i$  is censored ( $\delta_i = 0$ ) or not ( $\delta_i = 1$ ). The corresponding hazard function of lifetime is  $\lambda_F$  and cumulative hazard function is  $\Lambda_F$ .

## The variance comparison goal

We denote the estimate of  $F(t)$ :

- with Kaplan Meier as  $F_n^{km}(t)$
- with Dikta's method:  $\hat{F}_n^D(t)$
- with the new method:  $\hat{F}_n^N(t)$

We denote the asymptotic variance of  $n^{1/2}(F_n^{km}(t) - F(t))$  by  $v^{km}(t)$ , that of  $n^{1/2}(F_n^D(t) - F(t))$  by  $v^D(t)$ , that of  $n^{1/2}(F_n^N(t) - F(t))$  by  $v^N(t)$ .

We have know that  $v^{km}(t) > v^D(t)$  from the Dikta's paper. Our goal is to compare  $v^{km}(t)$  with  $v^N(t)$ , and  $v^D(t)$  with  $v^N(t)$

We have that

$$v^{km}(t) = (1 - F(t))^2 \int_0^t \frac{1}{(1 - H(x))^2} H^1(dx)$$

$$v^{km}(t) - v^D(t) = (1 - F(t))^2 r(t) \geq 0$$

$$v^D(t) = v^{km}(t) - (1 - F(t))^2 r(t) = (1 - F(t))^2 \left[ \int_0^t \frac{1}{(1 - H(x))^2} H^1(dx) - r(t) \right]$$

where

$$r(t) = \int_0^t \frac{1 - m(x, \theta_0)}{(1 - H(x))^2} H^1(dx) - \int_0^t \int_0^t \frac{\alpha(x, y)}{(1 - H(x))(1 - H(y))} H(dy) H(dx)$$

$$\alpha(x, y) = \langle \text{Grad}(m(x, \theta_0)), (I^{-1}(\theta_0) \text{Grad}(m(x, \theta_0))) \rangle$$

Our goal is to find

$$n^{1/2}(F_n^N(t) - F(t))$$

Since there is  $n^{1/2}(\Lambda_n(t) - \Lambda(t))$ , could we use a Delta method to calculate  $n^{1/2}(F_n^N(t) - F(t))$ ?

**New assumption:**

Instead of the strong assumption of independent between  $T_i$  and  $C_i$ , we proposed that  $T \perp\!\!\!\perp C$  at a small neighborhood, where  $T = C$ . That is, we have

$$\lim_{dt \rightarrow 0} P(C > t, T \geq t + dt) = P(C > t)P(T \geq t + dt) \quad (1)$$

As well as

$$P(C > t, T \geq t) = P(C > t)P(T \geq t) \quad (2)$$

With this assumption, we can show:

$$\begin{aligned} P(C > t|T = t) &= \lim_{dt \rightarrow 0} P(C > t|t \leq T < t + dt) \\ &= \lim_{dt \rightarrow 0} \frac{P(C > t, t \leq T < t + dt)}{P(t \leq T < t + dt)} \\ &= \lim_{dt \rightarrow 0} \frac{P(C > t, T \geq t) - P(C > t, T > t + dt)}{P(T \geq t) - P(T > t + dt)} \\ &= \lim_{dt \rightarrow 0} \frac{P(C > t)(P(T \geq t) - P(T > t + dt))}{P(T \geq t) - P(T > t + dt)} \\ &= P(C > t) \end{aligned} \quad (3)$$

And since independent,

$$P(C > t|T > t) = \frac{P(C > t, T > t)}{P(T > t)} = \frac{P(C > t)P(T > t)}{P(T > t)} = P(C > t)$$

Therefore,

$$P(C > t|T > t) = P(C > t|T = t) \quad (4)$$

Given (Eq 1), we could derive that

$$P(\delta = 1|X = t) = \frac{P(C > t, T = t)}{P(X = t)} = \frac{P(T = t)}{P(X = t)} \frac{P(C > t, T > t)}{P(T > t)} = \frac{f(t)S_x(t)}{h(t)S(t)} = \frac{\lambda_F(t)}{\lambda_H(t)}$$

where  $\lambda_H(t)$  is the hazard function corresponding to  $Z$ , which is known as crude hazard rate as well.

We may define  $m(t) = P(\delta = 1|X = t) = E(\delta|X = t)$ . Then

$$m(t) = \frac{\lambda_F(t)}{\lambda_H(t)} \quad (5)$$

which is the same parameter defined in Dikta's papers. Therefore, the independence between  $T$  and  $C$  is not the necessary condition for equation (2).

## A question about the $H^1, m(t)$ relationship

$$m(x) = E(\delta = 1|Z = x) = P(\delta = 1|Z = x) = \frac{\lambda_F(x)}{\lambda_H(x)}$$

And there is a relationship in Dikta's paper:

$$\begin{aligned} H_1(x) &= P(\delta = 1, Z \leq x) = P(T \leq x, T < C) \\ &= \int_0^x \int_t^\infty f_{ts}(t, s) ds dt \end{aligned} \quad (6)$$

There is another relationship in the paper

$$\begin{aligned} H_1(x) &= P(\delta = 1, Z \leq x) = \int_0^x \bar{G}(t) F(dt) \\ &= \int_0^x \int_t^\infty g(s) ds f(t) dt \\ &= \int_0^x \int_t^\infty f(t) g(s) ds dt \end{aligned} \quad (7)$$

And eq (6) = eq (7), is that means that  $f_{t,s}(t, s) = f(t)g(s)$  every where?

**$m(t)$  function,  $H_1(t)$  function**

$$H_1(x) = P(\delta = 1, Z \leq x) = \int_0^x m(z) H(dz)$$

Since  $m(x) = \frac{\lambda_F(x)}{\lambda_H(x)}$ , then  $\int_0^x m(z) H(dz) = \int_0^x \lambda_F(z) dz$ , then  $H_1(x) = P(\delta = 1, Z \leq x) = \int_0^x \lambda_F(z) dz$ ? is that correct?

$$H^1(t) = \int_0^t m(x) H(dx)$$

$$\begin{aligned} \Lambda(t) &= \int_0^t \frac{1}{1 - F(x)} F(dx) = \int_0^t \frac{1}{1 - H(x)} H^1(dx) \\ &\rightarrow \Lambda(t) = \int_0^t \frac{m(x)}{1 - H(x)} H(dx) \end{aligned}$$

## Formulas

### The relationship

$$m(x) = P(\delta = 1|Z = x) = E(\delta|Z = x)$$

$$m(x) = \frac{\lambda_F}{\lambda_H} = \frac{f}{S} \frac{S_h}{f_h}$$

$$H^1(t) = P(\delta = 1, Z \leq x) = \int_0^x m(z)h(z)dz$$

$$= \int_0^x \frac{f(z)}{S(z)} \frac{S_h(z)}{d} dz,$$

$$\text{since } S_h(x) = P(Z > x) = P(T > x)P(C > x) = S(x)S_c(x)$$

$$= \int_0^x f(z)S_c(z)dz = \int_0^x f(z)G(z)dz, \text{ (we denote } G(z) = S_c(z))$$

$$H^1(dt) = m(z)h(z)dz$$

$$\Lambda(t) = \int_0^t \frac{f(x)}{S(x)} dx = \int_0^t \lambda_f(x) dx = \int_0^t m(x)\lambda_H(x) dx = \int_0^t \frac{m(x)f_H(x)}{S_H(x)} dx = \int_0^t \frac{m(x)}{1 - H(x)} H(dx)$$

Since  $H^1(t) = \int_0^x m(z)h(z)dz$ ,

$$\Lambda(t) = \int_0^t \frac{m(x)}{1 - H(x)} H(dx) = \int_0^t \frac{1}{1 - H(x)} H^1(dx)$$