# Restricted Mean Survival Time

### 1.1 Setup

Let T denote the time to event and C a random censoring time. Define:

$$U = \min(T, C),$$
  $\delta = \begin{cases} 1 & \text{if observed,} \\ 0 & \text{if censored.} \end{cases}$ 

The **restricted mean survival time** (RMST)  $\theta(\tau)$  is the area under the survival curve  $S(t) = \mathbb{P}(T > t)$  up to time  $\tau$ :

$$\theta = \theta(\tau) = \int_0^{\tau} S(t)dt.$$

An estimator for  $\theta(\tau)$  is given by:

$$\theta(\tau) = \int_0^{\tau} \hat{S}(t)dt,$$

where  $\hat{S}(t)$  is the Kaplan-Meier (KM) estimator of the survival function.

## 1.2 Asymptotics

**Proposition 1.1.** Define  $\mu(t,\tau)$  as the area under the survival curve from time t to the truncation time  $\tau$ :

$$\mu(t,\tau) = \int_{t}^{\tau} S(u)du.$$

The standardized process  $\sqrt{n}\{\hat{\theta}(\tau) - \theta(\tau)\}$  converges weakly to a Brownian motion:

$$\sqrt{n} \{ \hat{\theta}(\tau) - \theta(\tau) \} \rightsquigarrow W \{ \sigma_{\text{RMST}}^2(\tau) \},$$

with variance function:

$$\sigma_{\text{RMST}}^2(\tau) = \int_0^{\tau} \frac{\mu^2(t,\tau)\alpha(t)}{y(t)} dt.$$

Here  $\alpha(t)$  is the (instantaneous) hazard at time t, and y(t) is the probability of being at risk at time t:

$$y(t) = \lim_{n \to \infty} n^{-1}Y(t) = \mathbb{P}(U > t).$$

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**Proof.** Note first that:

$$\frac{d\mu(t,\tau)}{dt} = -S(t)dt,$$

or  $d\mu(t,\tau) = -S(t)dt$ . Recall that the Kaplan-Meier and Nelson-Aalen estimators are asymptotically linked via:

$$\frac{\sqrt{n}\{\hat{S}(u) - S(u)\}}{-S(u)} = \sqrt{n}\{\hat{A}(u) - A(u)\} + o_p(1).$$

The standardized RMST process is:

$$\sqrt{n} \{ \hat{\theta}(\tau) - \theta(\tau) \} = \int_0^{\tau} \sqrt{n} \{ \hat{S}(u) - S(u) \} du$$

$$= \int_0^{\tau} \frac{\sqrt{n} \{ \hat{S}(u) - S(u) \}}{-S(u)} \cdot \{ -S(u) du \}$$

$$= \int_0^{\tau} \sqrt{n} \{ \hat{A}(u) - A(u) \} d\mu(t, \tau) + o_p(1).$$

Integrating by parts:

$$\int_{0}^{\tau} \sqrt{n} \{\hat{A}(u) - A(u)\} d\mu(t, \tau) = \underbrace{\left[\sqrt{n} \{\hat{A}(u) - A(u)\}\mu(t, \tau)\right]_{t=0}^{t=\tau} - \int_{0}^{\tau} \mu(t, \tau) \cdot \sqrt{n} d\{\hat{A}(u) - A(u)\}}_{t=0}.$$

The first term on the right vanishes because the cumulative hazard is zero at time zero  $\hat{A}(0) = A(0) = 0$ , and  $\mu(\tau, \tau) = 0$ . Thus:

$$\sqrt{n} \left\{ \hat{\theta}(\tau) - \theta(\tau) \right\} = -\int_0^\tau \mu(t, \tau) \cdot \sqrt{n} d\left\{ \hat{A}(u) - A(u) \right\} + o_p(1). \tag{1.1}$$

The standardized Nelson-Aalen process is expressible as a zero-mean martingale:

$$\sqrt{n} \{ \hat{A}(u) - A(u) \} = \int_0^u \frac{\sqrt{n}}{Y(t)} dM(t).$$

Here  $dM(u) = dN(u) - \alpha(u)Y(u)du$  is the increment in the counting process martingale for N(u), the **total** number of events by time u. Substituting

$$\sqrt{n}d\{\hat{A}(u) - A(u)\} = \frac{\sqrt{n}}{Y(u)}dM(u).$$

into (1.1) gives:

$$\sqrt{n} \{ \hat{\theta}(\tau) - \theta(\tau) \} = -\int_0^\tau \frac{\mu(u, \tau) \sqrt{n}}{Y(u)} dM(u) + o_p(1).$$

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Finally, taking the predictable variation:

$$\left\langle \sqrt{n} \left\{ \hat{\theta}(\tau) - \theta(\tau) \right\} \right\rangle = \int_0^{\tau} \frac{n\mu^2(t,\tau)}{Y^2(t)} d\langle M(t) \rangle + o_p(1)$$

$$= \int_0^{\tau} \frac{n\mu^2(t,\tau)}{Y^2(t)} Y(t) \alpha(t) dt + o_p(1)$$

$$= \int_0^{\tau} \frac{\mu^2(t,\tau)\alpha(t)}{n^{-1}Y(t)} dt + o_p(1)$$

$$\xrightarrow{p} \int_0^{\tau} \frac{\mu^2(t,\tau)\alpha(t)}{y(t)} dt,$$

#### 1.3 Influence Function

Taking  $\tau$  as fixed, the influence function expansion for the RMST is:

$$\sqrt{n}(\hat{\theta} - \theta) = \frac{1}{\sqrt{n}} \sum_{i=1}^{n} \psi_i + o_p(1),$$

$$\psi_i = -\int_0^{\tau} \frac{\mu(t, \tau)}{y(t)} dM(t)$$
(1.2)

Here  $\mu(t,\tau) = \int_t^\tau S(u) du$ ,  $y(u) = \mathbb{P}(U > t)$ , and:

$$dM(t) = dN(t) - Y(t)dA(t) = \sum_{i=1}^{n} dN_i(t) - Y_i(t)dA(t).$$

The influence function contributions can be estimated by:

$$\hat{\psi}_i = -\int_0^{\tau} \frac{\hat{\mu}(t,\tau)}{n^{-1}Y(t)} \left\{ dN_i(t) - Y_i(t)d\hat{A}(t) \right\},$$

where  $\hat{\mu}(t,\tau) = \int_0^{\tau} \hat{S}(t)dt$  is the integral of the Kaplan-Meier curve  $\hat{S}(t)$ , and  $d\hat{A}(t)$  is the increment of the Nelson-Aalen estimator:

$$d\hat{A}(t) = \frac{\sum_{i=1}^{n} dN_i(t)}{Y(t)}$$

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1.4 Perturbation

The variance of  $\hat{\theta}(\tau)$  can be estimated analytically via:

$$\hat{\mathbb{V}}\{\hat{\theta}(\tau)\} = \hat{\sigma}_{\text{RMST}}^2(\tau) = \int_0^{\tau} \frac{\hat{\mu}^2(t,\tau)}{n^{-1}Y(t)} d\hat{A}(t).$$

Perturbation provides an alternative applicable to more complex estimators. To estimate the sampling variance via perturbation:

- 1. Precompute the influence function contributions  $\psi_i$  for all subjects.
- 2. For B iterations, repeat the following:
  - i. Draw n perturbation weights  $w_i^{(b)} \overset{\text{IID}}{\sim} N(0, 1)$ .
  - ii. Calculate and store:

$$D_b = \frac{1}{n} \sum_{i=1}^{n} w_i^{(b)} \psi_i.$$

3. Estimate  $\mathbb{V}(\hat{\theta})$  by taking the variance of the B realizations  $\{D_1, \dots, D_B\}$ .

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