

Exercise 4.3

The counting processes $N_1(t), \dots, N_u(t)$ have intensity processes $\lambda_i(t) = \gamma_i(t) \beta_0(t); i=1, \dots, u$.

Here $\gamma_i(t) = 0, 1$ is an at risk indicator.

This is a special case of the additive regression model with def. page 157 in ASAJ

$$N(t) = (N_1(t), \dots, N_u(t)), \quad B(t) = \beta_0(t) = \int_0^t \beta_0(u) du,$$

and $X(t) = (\gamma_1(t), \dots, \gamma_u(t))^T$. We note that (since $\gamma_i(t)$ only takes the values 0 and 1):

$$X(t)^T X(t) = \sum_{i=1}^u \gamma_i(t)^2 = \sum_{i=1}^u \gamma_i(t) = \gamma_0(t)$$

Further $J(t) = I(\text{rank } X(t) = 1) = I(\gamma_0(t) > 0)$.

Hence the estimator (4.59) takes the form

$$\begin{aligned} \hat{\beta}_0(t) &= \int_0^t J(u) (X(u)^T X(u))^{-1} X(u)^T dN(u) \\ &= \int_0^t J(u) \gamma_0(u)^{-1} \left(\sum_{i=1}^u \gamma_i(u) dN_i(u) \right) \\ &= \int_0^t (J(u) / \gamma_0(u)) dN_0(u) \end{aligned}$$

which is the Nelson-Halen estimator

Exercise 4.4

②

$N_i(t)$ has intensity process

$$\lambda_i(t) = Y_i(t) \{ \beta_0(t) + \beta_1(t) x_i \}, \text{ where } Y_i(t)$$

is an at risk indicator and $x_i = 0, 1$

is a group indicator. The number of

risk in group 0 (i.e. $x_i = 0$) is

$$Y^{(0)}(t) = \sum_{i=1}^n (1 - x_i) Y_i(t) \text{ and the number}$$

of risk in group 1 (i.e. $x_i = 1$) is

$$Y^{(1)}(t) = \sum_{i=1}^n x_i Y_i(t). \text{ Thus } Y_0(t) = Y^{(0)}(t) + Y^{(1)}(t)$$

$$= \sum_{i=1}^n Y_i(t).$$

a) We have that

$$X(t) = \begin{pmatrix} Y_1(t) & Y_1(t)x_1 \\ Y_2(t) & Y_2(t)x_2 \\ \vdots & \vdots \\ Y_n(t) & Y_n(t)x_n \end{pmatrix}$$

By a direct computation we get

(3)

$$\begin{aligned}
 X(t)^T X(t) &= \begin{pmatrix} \sum_{i=1}^n y_i(t)^2 & \sum_{i=1}^n y_i(t)^2 x_i \\ \sum_{i=1}^n y_i(t)^2 x_i & \sum_{i=1}^n y_i(t)^2 x_i^2 \end{pmatrix} \\
 &= \begin{pmatrix} \gamma_0(t) & \gamma^{(1)}(t) \\ \gamma^{(1)}(t) & \gamma^{(2)}(t) \end{pmatrix}
 \end{aligned}$$

where the last equality follows since $y_i(t)^2 = \gamma_i(t)$ and $x_i^2 = x_i$.

b) We have that

$$\begin{aligned}
 \det(X(t)^T X(t)) &= \gamma_0(t) \gamma^{(2)}(t) - \gamma^{(1)}(t)^2 \\
 &= \gamma^{(2)}(t) (\gamma_0(t) - \gamma^{(1)}(t)) \\
 &= \gamma^{(2)}(t) \gamma^{(1)}(t)
 \end{aligned}$$

It follows that (when

$\gamma^{(2)}(t) > 0$ and $\gamma^{(1)}(t) > 0$):

(4)

$$\begin{aligned}
 (X(t)^T X(t))^{-1} &= \frac{1}{\text{Det}(X(t)^T X(t))} \begin{pmatrix} \gamma^{(1)}(t) & -\gamma^{(2)}(t) \\ -\gamma^{(2)}(t) & \gamma_0(t) \end{pmatrix} \\
 &= \frac{1}{\gamma^{(2)}(t)\gamma^{(1)}(t)} \begin{pmatrix} \gamma^{(1)}(t) & -\gamma^{(2)}(t) \\ -\gamma^{(2)}(t) & \gamma^{(0)}(t) + \gamma^{(1)}(t) \end{pmatrix} \\
 &= \begin{pmatrix} \frac{1}{\gamma^{(2)}(t)} & -\frac{1}{\gamma^{(2)}(t)} \\ -\frac{1}{\gamma^{(2)}(t)} & \frac{1}{\gamma^{(0)}(t)} + \frac{1}{\gamma^{(1)}(t)} \end{pmatrix}
 \end{aligned}$$

9) The estimator $\hat{B}(t) = (\hat{B}_0(t), \hat{B}_1(t))^T$ is given by [cf. (4.58) & (4.59) in ABG]:

$$\hat{B}(t) = \int_0^t J(u) (X(u)^T X(u))^{-1} X(u)^T dN(u)$$

where $N(t) = (N_1(t), \dots, N_n(t))^T$.

(5)

Now we have that

$$(X(t)^T X(t))^{-1} X(t)^T$$

$$= \begin{pmatrix} \frac{y_1(t)}{y^{(0)}(t)} - \frac{y_1(t)x_1}{y^{(0)}(t)} & \dots & \frac{y_n(t)}{y^{(0)}(t)} - \frac{y_n(t)x_n}{y^{(0)}(t)} \\ \frac{y_1(t)x_1}{y^{(0)}(t)} + \frac{y_1(t)x_1}{y^{(1)}(t)} - \frac{y_1(t)}{y^{(0)}(t)} & \dots & \frac{y_n(t)x_n}{y^{(0)}(t)} + \frac{y_n(t)x_n}{y^{(1)}(t)} - \frac{y_n(t)}{y^{(0)}(t)} \end{pmatrix}$$

Hence we get

$$\hat{\beta}(t) = \int_0^t J(u) \begin{pmatrix} \sum_{i=1}^n \frac{(1-x_i)y_i(u)}{y^{(0)}(u)} dN_i(u) \\ \sum_{i=1}^n \frac{x_i y_i(u)}{y^{(1)}(u)} dN_i(u) - \sum_{i=1}^n \frac{(1-x_i)y_i(u)}{y^{(0)}(u)} dN_i(u) \end{pmatrix}$$

$$= \int_0^t J(u) \begin{pmatrix} dN^{(0)}(u)/y^{(0)}(u) \\ dN^{(1)}(u)/y^{(1)}(u) - dN^{(0)}(u)/y^{(0)}(u) \end{pmatrix}$$

Where $N^{(0)}(t) = \sum_{i=1}^n (1-x_i) N_i(t)$ and $N^{(1)}(t) = \sum_{i=1}^n x_i N_i(t)$

Thus $\hat{\beta}_0(t) = \int_0^t (J(u)/y^{(0)}(u)) dN^{(0)}(u)$ is the Nelson-Aalen estimator for group 0 and $\hat{\beta}_1(t) = \int_0^t (J(u)/y^{(1)}(u)) dN^{(1)}(u) - \int_0^t (J(u)/y^{(0)}(u)) dN^{(0)}(u)$ is the difference between the Nelson-Aalen estimators on the two groups.

(6)

Exercise 4.5

We consider the situation of exercise 4.4

We want to test the null hypothesis

$H_0: \beta_1(t) = 0$ for $0 \leq t \leq t_0$. The TST

test statistic is given by

$$U_1(t_0) = \frac{Z_1(t_0)}{\sqrt{V_{11}(t_0)}}$$

where $Z_1(t_0)$ and $V_{11}(t_0)$ are given by (4.66) and (4.67) in the ABG-book, where the weight process $L_1(t)$ is given as described below (4.68).

a) By result b) in exercise 4.4, we have

$$(X(t)^T X(t))^{-1} = \begin{pmatrix} \frac{1}{\gamma^{(0)}(t)} & -\frac{1}{\gamma^{(0)}(t)} \\ -\frac{1}{\gamma^{(0)}(t)} & \frac{1}{\gamma^{(0)}(t)} + \frac{1}{\gamma^{(1)}(t)} \end{pmatrix}$$

From this we obtain

$$\text{diag} \left[(X(t)^T X(t))^{-1} \right] = \begin{pmatrix} \frac{1}{Y^{(0)}(t)} & 0 \\ 0 & \frac{1}{Y^{(0)}(t)} + \frac{1}{Y^{(1)}(t)} \end{pmatrix} \quad (7)$$

and hence

$$K(t) = \left\{ \text{diag} \left[(X(t)^T X(t))^{-1} \right] \right\}^{-1}$$

$$= \begin{pmatrix} Y^{(0)}(t) & 0 \\ 0 & \frac{Y^{(0)}(t) Y^{(1)}(t)}{Y_0(t)} \end{pmatrix}$$

The weight process $L_1(t)$ is given as $K_{11}(t)$, i.e. the element in the lower right-hand corner of $K(t)$.

Thus

$$L_1(t) = \frac{Y^{(0)}(t) Y^{(1)}(t)}{Y_0(t)}$$

⑧

Using (4.66) and the result in question c of exercise 4.4, we find that the test statistic $Z_1(t_0)$ takes the form

$$\begin{aligned} Z_1(t_0) &= \int_0^{t_0} L_1(t) d\tilde{B}_1(t) \\ &= \int_0^{t_0} \frac{Y^{(0)}(t) Y^{(1)}(t)}{Y_0(t)} \left(\frac{dN^{(1)}(t)}{Y^{(1)}(t)} - \frac{dN^{(0)}(t)}{Y^{(0)}(t)} \right) \end{aligned}$$

which equals (3.52) when we use the log-rank weights (cf table 3.2 on page 107 in ASG)

b) The variance estimator is given as [cf (4.67)]

$$\begin{aligned} V_{11}(t_0) &= \int_0^{t_0} L_1(t)^2 d\tilde{\sigma}_{11}(t) \\ &= \sum_{T_j \leq t_0} L_1(T_j)^2 \Delta \tilde{\sigma}_{11}(T_j) \end{aligned}$$

⑨

As described on page 164 in ABG
there are two options for computing
 $\hat{\sigma}_n(t)$. We may either use
(4.63), which is obtained from
(4.62) by replacing $\lambda(t)$ in
(4.62) by $dN(t)$. Or we may
use (4.64), which is obtained
by replacing $\lambda(t)$ by the
estimated intensity process under H_0 .

Then (4.64) takes the form

$$\Sigma_{\text{mod}}(t) = \sum_{T_j \leq t} J(T_j) X^{-}(T_j) \text{diag} \{ X_0(T_j) \Delta \hat{B}_0(T_j) \} X^{-}(T_j)^T \quad (\#)$$

where $X^{-}(t) = (X(t)^T X(t))^{-1} X(t)^T$

is defined for the model of

exercise 4.4, and an expression

(10)

is given on the top of page 5
in the solution of exercise 4.4.

Further $X_0(t)$ and $\int B_0(t)$ are given
under H_0 , which corresponds to
the situation in exercise 4.3.

Thus we have

$$X_0(t) = (Y_1(t), \dots, Y_n(t))^T$$

$$\text{and } \int B_0(t) = \int_0^t \frac{dN_0(u)}{Y_0(u)}$$

By a direct matrix multiplication
we then find after (quite some) algebra that

$$X^{-1}(T_j) \text{diag} \{ X_0(T_j) \Delta B_0(T_j) \} X^{-1}(T_j)^T$$

$$= \begin{pmatrix} \frac{1}{Y^{(0)}(T_j)} \frac{\Delta N_0(T_j)}{Y_0(T_j)} & -\frac{1}{Y^{(0)}(T_j)} \frac{\Delta N_0(T_j)}{Y_0(T_j)} \\ -\frac{1}{Y^{(0)}(T_j)} \frac{\Delta N_0(T_j)}{Y_0(T_j)} & \left(\frac{1}{Y^{(0)}(T_j)} + \frac{1}{Y^{(0)}(T_j)} \right) \frac{\Delta N_0(T_j)}{Y_0(T_j)} \end{pmatrix}$$

①

Further we have that $J(T_j)$ in (#) is equal to 1 if there is at least one individual at risk in each group.

From this we obtain

$$\begin{aligned} \Delta \hat{\sigma}_u(T_j) &= J(T_j) \left(\frac{1}{Y^{(1)}(T_j)} + \frac{1}{Y^{(0)}(T_j)} \right) \frac{\Delta N_o(T_j)}{Y_o(T_j)} \\ &= J(T_j) \frac{Y_o(T_j)}{Y^{(1)}(T_j) Y^{(0)}(T_j)} \frac{\Delta N_o(T_j)}{Y_o(T_j)} \end{aligned}$$

and hence the variance estimator of the test statistic becomes

$$V_u(t_0) = \sum_{T_j \leq t_0} L_1(T_j)^2 \Delta \hat{\sigma}_u(T_j)$$

$$= \sum_{T_j \leq t_0} \left(\frac{Y^{(0)}(T_j) Y^{(1)}(T_j)}{Y_o(T_j)} \right)^2 \frac{Y_o(T_j)}{Y^{(1)}(T_j) Y^{(0)}(T_j)} \frac{\Delta N_o(T_j)}{Y_o(T_j)}$$

$$= \sum_{T_j \leq t_0} \frac{Y^{(0)}(T_j) Y^{(1)}(T_j)}{Y_0(T_j)} \Delta N_0(T_j) \quad (\# \#)$$

The variance estimator of the logrank test is given by [cf (3.55)]

$$\begin{aligned} & \int_0^{t_0} \frac{L^2(t)}{Y_1(t) Y_2(t)} dN_0(t) \\ &= \int_0^{t_0} \left(\frac{Y_1(t) Y_2(t)}{Y_0(t)} \right)^2 \frac{dN_0(t)}{Y_1(t) Y_2(t)} \\ &= \int_0^{t_0} \frac{Y_1(t) Y_2(t)}{Y_0(t)^2} dN_0(t) \end{aligned}$$

which is the same as (# #).