

SOLUTION TO EXERCISES WEEK 39

Exercise 3.5

X_1, X_2, \dots, X_n are iid nonnegative random variables with survival function $S(t) = P(X_i > t)$ and cumulative distribution function $F(t) = P(X_i \leq t) = 1 - S(t)$.

a) When there are no censored observations, the number at risk $Y(t)$ takes the form

$$Y(t) = \sum_{i=1}^n I(X_i > t) = n - \sum_{i=1}^n I(X_i \leq t)$$

In particular, if we let $X_{(1)} < X_{(2)} < \dots < X_{(n)}$ be the ordered X_i 's, we have

$$Y(X_{(1)}) = n$$

$$Y(X_{(2)}) = n - 1$$

$$Y(X_{(3)}) = n - 2,$$

etc

i.e $Y(X_{(j)}) = n - j + 1$

(2)

For $X_{(j)} \leq t < X_{(j+1)}$ we therefore have

$$\hat{S}(t) = \prod_{X_{(i)} \leq t} \left(1 - \frac{1}{Y(X_{(i)})} \right)$$

$$= \prod_{X_{(i)} \leq t} \frac{Y(X_{(i)}) - 1}{Y(X_{(i)})}$$

$$= \frac{n-1}{n} \cdot \frac{n-2}{n-1} \cdot \frac{n-3}{n-2} \cdots \frac{n-j+1}{n-j+2} \cdot \frac{n-j}{n-j+1}$$

$$= \frac{n-j}{n} = 1 - \frac{j}{n}$$

$$= 1 - \hat{F}(t)$$

Since $\hat{F}(t) = n^{-1} \sum_{i=1}^n I(X_i \leq t) = j/n$

for $X_{(j)} \leq t < X_{(j+1)}$.

(3)

b) Greenwood's formula takes the form

$$\tilde{\tau}^2(t) = \hat{S}(t)^2 \sum_{X_{(i)} \leq t} \frac{1}{Y(X_{(i)}) (Y(X_{(i)}) - 1)}$$

In order to show that $\tilde{\tau}^2(t) = \frac{\hat{S}(t)(1-\hat{S}(t))}{n}$, it is sufficient to show that

$$\frac{\tilde{\tau}^2(t)}{\hat{S}(t)^2} = \sum_{X_{(i)} \leq t} \frac{1}{Y(X_{(i)}) (Y(X_{(i)}) - 1)} \quad (*)$$

is equal to $(1 - \hat{S}(t)) / (n \hat{S}(t))$.

We will show this by noting that both expressions equal zero when $t=0$, by noting that both expressions are constant between the $X_{(i)}$'s, and by showing that the increments at the $X_{(i)}$'s are the same.

The increment of (*) at $X_{(j)}$
equals

$$\frac{1}{Y(X_{(j)}) (Y(X_{(j)}) - 1)} = \frac{1}{(n-j+1)(n-j)}$$

While the increment of $(1 - \hat{S}(t)) / (n \hat{S}(t))$
at $t = X_{(j)}$ equals

$$\begin{aligned} & \frac{1 - \hat{S}(X_{(j)})}{n \hat{S}(X_{(j)})} - \frac{1 - \hat{S}(X_{(j-1)})}{n \hat{S}(X_{(j-1)})} \\ &= \frac{1 - (1 - \frac{j}{n})}{n (1 - \frac{j}{n})} - \frac{1 - (1 - \frac{j-1}{n})}{n (1 - \frac{j-1}{n})} \\ &= \frac{j}{n(n-j)} - \frac{j-1}{n(n-j+1)} \\ &= \frac{j(n-j+1) - (j-1)(n-j)}{n(n-j+1)(n-j)} = \frac{1}{(n-j+1)(n-j)} \end{aligned}$$

This shows the result.

Exercise 3.6

We have that $\hat{S}(t)$ is approximately normally distributed with mean $S(t)$ and a variance that may be estimated by $\hat{v}^2(t)$.

By an argument similar to the one in exercise 3.3, we have that

$$g(\hat{S}(t)) \pm z_{1-\alpha/2} g'(\hat{S}(t)) \hat{v}^2(t)$$

is an approximate 100(1- α)% confidence interval for $g(S(t))$. Here $g(x)$ is a strictly increasing continuously differentiable function.

For $g(x) = -\log(-\log x)$, we have

$$g'(x) = -1/(x \log x). \quad [\text{Note that}$$

$$g'(x) > 0 \quad \text{for } x \in (0, 1).]$$

Thus a $100(1-\alpha)\%$ confidence interval for $-\log(-\log S(t))$ is

("±" means + for the upper limit and minus for the lower limit)

$$-\log(-\log \hat{S}(t)) \pm z_{1-\alpha/2} \left(\frac{-1}{\hat{S}(t) \log \hat{S}(t)} \right) \hat{\tau}(t)$$

It follows that a confidence interval for $\exp(-\log(-\log S(t))) = -1/\log S(t)$ is given by

$$\frac{-1}{\log \hat{S}(t)} \exp \left(\pm z_{1-\alpha/2} \frac{-\hat{\tau}(t)}{\hat{S}(t) \log \hat{S}(t)} \right)$$

Hence an interval for $\log S(t)$ is (note that $\log S(t) < 0$):

$$\log \hat{S}(t) \cdot \exp \left(\pm z_{1-\alpha/2} \frac{\hat{\tau}(t)}{\hat{S}(t) \log \hat{S}(t)} \right)$$

Finally this yields the following interval for $S(t)$

$$\hat{S}(t) \exp(\pm z_{1-\alpha/2} \hat{\sigma}(t) / (\hat{S}(t) \log \hat{S}(t)))$$

cf. (3-30) in ABG.

Exercise 3.8

We have that $(\hat{S}(t) - S(t)) / \hat{\sigma}(t)$ is approximately standard normally distributed

a) Let ξ_p be the p -th fractile of the survival distribution. Then $S(\xi_p) = 1 - p$. In order to test the hypothesis

$$H_0: \xi_p = \xi_p^0 \quad \text{vs} \quad H_A: \xi_p \neq \xi_p^0$$

We may consider the test statistic

$$Z = \frac{\hat{S}(\xi_p^0) - S(\xi_p^0)}{\hat{v}(\xi_p^0)} = \frac{\hat{S}(\xi_p^0) - (1-p)}{\hat{v}(\xi_p^0)}$$

which is approximately $N(0,1)$ distributed under H_0 . We may therefore reject H_0 if

$$|Z| > z_{1-\alpha/2}$$

b) According to a general result, we get a $100(1-\alpha)\%$ confidence for ξ_p as all ξ_p^0 -values that are not rejected by the test in a.

This confidence interval is given by

$$\begin{aligned} & \left\{ \xi_p^0 : \left| \hat{S}(\xi_p^0) - (1-p) \right| / \hat{v}(\xi_p^0) \leq z_{1-\alpha/2} \right\} \\ & = \left\{ \xi_p^0 : \left| \hat{S}(\xi_p^0) - (1-p) \right| \leq z_{1-\alpha/2} \hat{v}(\xi_p^0) \right\} \end{aligned}$$

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The confidence interval may be read off a plot of the Kaplan-Meier estimator with confidence limits as illustrated in Fig. 3.13 in ABH.

An explanation is given in the figure below:

