



Construction of Actuarial Models

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Solutions to practice questions - Chapter 8

Solution 8.1

From the given baseline hazard function, we have:

$$H_0(x) = \int_0^x \frac{2}{10+s} ds = 2\ln(10+s)\Big|_0^x = 2\ln\left(\frac{10+x}{10}\right) = \ln\left(\left(\frac{10+x}{10}\right)^2\right)$$

$$\Rightarrow S_0(x) = e^{-H_0(x)} = \left(\frac{10}{10+x}\right)^2$$

$$\Rightarrow S_1(x) = \frac{S_0(2+x)}{S_0(2)} = \left(\frac{10}{10+(x+2)}\right)^2 / \left(\frac{10}{10+2}\right)^2 = \left(\frac{12}{12+x}\right)^2$$

Solution 8.2

$$\begin{split} h_0\left(x\right) &= 1.2 \, / \big(100 - x\big) \implies H_0\left(x\right) = 1.2 \ln \left(\frac{100}{100 - x}\right) \\ &\Rightarrow S_0\left(x\right) = e^{-H_0(x)} = \left(\frac{100}{100 - x}\right)^{-1.2} = \left(\frac{100 - x}{100}\right)^{1.2} \quad \text{for } 0 \le x < 100 \\ &\Rightarrow S_1\left(x\right) = S\left(1.05x\right) = \left(\frac{100 - 1.05x}{100}\right)^{1.2} = \left(\frac{95.23810 - x}{95.23810}\right)^{1.2} \quad \text{for } 0 \le x < 95.23810 \end{split}$$

Solution 8.3

In general, if hazard functions are related by $h_1(x) = k h_0(x)$, then survival functions are related by $S_1(x) = (S_0(x))^k$. So here we have:

$$S_1(x) = \sqrt{\frac{10}{10 + x}} \quad \text{for } x > 0$$

The hazard function here is the one corresponding to the baseline survival function given in Question 8.3:

$$h_1(x) = 0.5 h_0(x) = \frac{0.5}{10 + x}$$
 $\Rightarrow S_1(x) = \sqrt{\frac{10}{10 + x}} \Rightarrow f_1(x) = h_1(x) S_1(x) = \frac{0.5\sqrt{10}}{(10 + x)^{1.5}}$

Solution 8.5

$$H_0(x) = \int_0^x 0.002t \, dt = 0.001x^2 \text{ for } x > 0 \implies S_0(x) = e^{-0.001x^2} \text{ for } x > 0$$

$$\implies {}_{10} p_{20} = \frac{S_0(30)}{S_0(20)} = \frac{e^{-0.001 \times 30^2}}{e^{-0.001 \times 20^2}} = e^{-0.5} = 0.60653$$

$$h_1(x) = 0.95 h_0(x) \implies {}_{10} p_{20} = 0.60653^{0.95} = 0.62189$$

Solution 8.6

$$S_i(x) = S_j(x)^{K_{ij}} \implies 0.82 = S_i(60) = S_j(60)^{K_{ij}} = 0.80^{K_{ij}} \implies K_{ij} = \frac{\ln(0.82)}{\ln(0.80)} = 0.889$$

Solution 8.7

$$H_{0}(x) = \int_{0}^{x} h_{0}(t)dt = \int_{0}^{x} 0.015 + 0.01 \times 1.01^{t} dt = 0.015x + \frac{0.01(1.01^{x} - 1)}{\ln(1.01)} \Rightarrow 10p_{60}^{\text{fn}} = \frac{S_{0}(70)}{S_{0}(60)} = \frac{\exp(-H_{0}(70))}{\exp(-H_{0}(60))} = \exp(-(H_{0}(70) - H_{0}(60)))$$

$$= \exp\left(-\left(0.015(70 - 60) + \frac{0.01(1.01^{70} - 1.01^{60})}{\ln(1.01)}\right)\right) = 0.71105$$

Solution 8.8

A female non-smoker ($z_1 = z_2 = 0$) corresponds to the baseline group. For a male 2-pack per day smoker ($z_1 = 1, z_2 = 2$) we have:

$$g(1,2) = e^{0.010 \times 1 + 0.015 \times 2} = e^{0.04} = 1.04081$$

So the relative risk is:

$$K = \frac{g(1,2)}{g(0,0)} = g(1,2) = 1.04081$$

Using the results of Solutions 8.8 and 8.9, we have:

$$_{10}p_{60}^{\text{ms}} = \left(_{10}p_{60}^{\text{fn}}\right)^{K} = 0.71105^{1.04081} = 0.70122$$

Solution 8.10

From Section 8.4, we have:

$$\frac{\partial \ln(L(K,\theta))}{\partial K} = \frac{t}{K} - \frac{\sum x_{m,j}}{\theta}$$

$$\frac{\partial \ln(L(K,\theta))}{\partial \theta} = \frac{\sum x_{f,i} + K \sum x_{m,j}}{\theta^2} - \frac{s+t}{\theta}$$

The second order partial derivatives are thus:

$$\frac{\partial^{2} \ln(L(K,\theta))}{\partial K^{2}} = -\frac{t}{K^{2}}$$

$$\frac{\partial^{2} \ln(L(K,\theta))}{\partial K \partial \theta} = \frac{\sum x_{m,j}}{\theta^{2}}$$

$$\frac{\partial^{2} \ln(L(K,\theta))}{\partial K^{2}} = -2 \times \frac{\sum x_{f,i} + K \sum x_{m,j}}{\theta^{3}} + \frac{s+t}{\theta^{2}}$$

Using the fact that $E[X_f] = \theta$ and $E[X_m] = \theta / K$ (see Section 8.4), we have:

$$-E\left[\frac{\partial^{2} \ln(L(K,\theta))}{\partial K^{2}}\right] = \frac{t}{K^{2}}$$

$$-E\left[\frac{\partial^{2} \ln(L(K,\theta))}{\partial K \partial \theta}\right] = -E\left[\frac{\sum X_{m,j}}{\theta^{2}}\right] = \frac{-t\theta/K}{\theta^{2}} = -\frac{t}{\theta K}$$

$$-E\left[\frac{\partial^{2} \ln(L(K,\theta))}{\partial K^{2}}\right] = 2 \times E\left[\frac{\sum X_{f,i} + K \sum X_{m,j}}{\theta^{3}}\right] - \frac{s+t}{\theta^{2}}$$

$$= 2 \times \frac{s\theta + tK(\theta/K)}{\theta^{3}} - \frac{s+t}{\theta^{2}} = \frac{s+t}{\theta^{2}}$$

So the information matrix is:

$$I(K,\theta) = \begin{pmatrix} \frac{t}{K^2} & -\frac{t}{\theta K} \\ -\frac{t}{\theta K} & \frac{s+t}{\theta^2} \end{pmatrix}$$

The asymptotic covariance matrix of the estimators is the inverse of the information matrix:

$$\begin{pmatrix} \operatorname{var}(\hat{K}) & \operatorname{cov}(\hat{K}, \hat{\theta}) \\ \operatorname{cov}(\hat{K}, \hat{\theta}) & \operatorname{var}(\hat{\theta}) \end{pmatrix} = I(K, \theta)^{-1} = \begin{pmatrix} \frac{t}{K^2} & -\frac{t}{\theta K} \\ -\frac{t}{\theta K} & \frac{s+t}{\theta^2} \end{pmatrix}^{-1}$$

$$= \frac{1}{\det(I(K, \theta))} \begin{pmatrix} \frac{s+t}{\theta^2} & \frac{t}{\theta K} \\ \frac{t}{\theta K} & \frac{t}{K^2} \end{pmatrix} = \frac{K^2 \theta^2}{st} \begin{pmatrix} \frac{s+t}{\theta^2} & \frac{t}{\theta K} \\ \frac{t}{\theta K} & \frac{t}{K^2} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{K^2(s+t)}{st} & \frac{K\theta}{s} \\ \frac{K\theta}{s} & \frac{\theta^2}{s} \end{pmatrix}$$

Solution 8.12

First determine the baseline (female) model:

$$h_0 = \frac{\alpha}{x} \quad \text{for } x \ge 1$$

$$H_0(x) = \int_1^x \frac{\alpha}{t} dt = \alpha \ln(x) \quad \text{for } x \ge 1$$

$$S_0(x) = e^{-H_0(x)} = e^{-\alpha \ln(x)} = \frac{1}{x^{\alpha}} \quad \text{for } x \ge 1$$

$$f_0(x) = h_0(x) S_0(x) = \frac{\alpha}{x^{\alpha+1}} \quad \text{for } x \ge 1$$

For the male survival model we will simplify the notation. Instead of the standard notation $h(x \mid z = 1) = g(1)h_0(x)$ we will simply write $h_1(x)$. Furthermore, it will simplify the likelihood function if we write $K = g(1) = e^{\beta}$:

$$h_1(x) = \frac{\alpha K}{x} \quad \text{for } x \ge 1$$

$$H_1(x) = \int_0^x h_1(t) dt = \alpha K \ln(x) \text{ for } x \ge 1$$

$$S_1(x) = e^{-H_1(x)} = \frac{1}{x^{\alpha K}} \text{ for } x \ge 1$$

$$f_1(x) = h_1(x) S_1(x) = \frac{\alpha K}{x^{\alpha K + 1}} \text{ for } x \ge 1$$

Now we can write down the likelihood function:

$$L(K,\alpha) = \prod_{i=1}^{s} f_0(x_{f,i}) \times \prod_{j=1}^{t} f_1(x_{m,j})$$

$$= \prod_{i=1}^{s} \frac{\alpha}{(x_{f,i})^{\alpha+1}} \times \prod_{j=1}^{t} \frac{\alpha K}{(x_{m,j})^{\alpha K+1}}$$

$$= \alpha^{s+t} K^t \times \frac{1}{\prod_{i=1}^{s} (x_{f,i})^{\alpha+1}} \times \frac{1}{\prod_{j=1}^{t} (x_{m,j})^{\alpha K+1}}$$

Solution 8.13

The log-likelihood function is:

$$\ln(L(K,\alpha)) = (s+t)\ln(\alpha) + t\ln(K) - (\alpha+1)\sum\ln(x_{f,i}) - (\alpha K+1)\sum\ln(x_{m,j})$$

The partial derivatives are:

$$\frac{\partial \ln(L(K,\alpha))}{\partial K} = \frac{t}{K} - \alpha \sum \ln(x_{m,j})$$

$$\frac{\partial \ln(L(K,\alpha))}{\partial \alpha} = \frac{s+t}{\alpha} - \sum \ln(x_{f,i}) - K \sum \ln(x_{m,j})$$

The simultaneous solutions are:

$$\hat{\alpha} = \frac{s}{\sum \ln(x_{f,i})}$$
 (inverse of the sample mean of the log of female lifetimes)
$$\hat{K} = \frac{\sum \ln(x_{f,i})/s}{\sum \ln(x_{m,j})/t}$$
 (ratio of sample means of log-lifetimes)

Solution 8.14

To simplify matters we will use the following notation:

female:
$$g(0) = e^{\beta \times 0} = 1$$

male: $g(1) = e^{\beta \times 1} = e^{\beta} = K$

The modified risk set data is:

Using the formula for the partial likelihood function and the notation introduced here, the partial likelihood function is:

$$L(K) = \prod_{i=1}^{4} \frac{K_i}{\sum_{j \text{ in } R_i} K_j}$$

$$= \frac{K}{2+2K} \times \frac{1}{2+K} \times \frac{1}{1+K} \times \frac{K}{1+K}$$

$$\underset{\text{male death}}{\underbrace{K_1: 2f, 2m}} \times \underbrace{\frac{1}{2+K}}_{R_2: 2f, m} \times \underbrace{\frac{1}{1+K}}_{R_3: 1f, 1m} \times \underbrace{\frac{K}{R_4: 1m}}_{\text{male death}}$$

This expression can be simplified to the following:

$$L(K) = \frac{K}{2(1+K)^2(2+K)}$$

The log-partial likelihood function is:

$$ln(L(K)) = ln(K) - 2ln(1+K) - ln(2+K) - ln(2)$$

It is maximized at the unique critical point:

$$0 = \frac{d \ln(L(K))}{dK} = \frac{1}{K} - \frac{2}{1+K} - \frac{1}{2+K}$$

$$\Rightarrow 0 = (1+K)(2+K) - 2K(2+K) - K(1+K) \text{ (numerator)}$$

$$\Rightarrow 0 = 1 - K - K^2$$

$$\Rightarrow \hat{K} = \frac{1 \pm \sqrt{5}}{-2} = 0.618 \text{ or } -1.618 \text{ (impossible since } K > 0)$$

$$\Rightarrow \hat{\beta} = \ln(\hat{K}) = -0.481$$

Solution 8.15

We obtained the partial likelihood estimate $\hat{K} = 0.618$ in Solution 8.14. We will need the following:

$$i 1 2 3 4$$

$$y_i 2 5 8 9$$

$$\frac{s_i}{\sum_{j \text{ in } R_i} K_j} \frac{1}{2+2K} \frac{1}{2+K} \frac{1}{1+K} \frac{1}{K}$$

$$= 0.309 0.382 0.618 1.618$$

Using the (unrounded) values in the last line in the display above leads to:

$$\hat{H}_{0}(x) = \sum_{y_{i} \leq x} \frac{s_{i}}{\sum_{j \text{ in } R_{i}}^{S_{i}}} = \begin{cases} 0 & 0 \leq x < 2\\ 0.309 & 2 \leq x < 5\\ 0.691 & 5 \leq x < 8\\ 1.309 & 8 \leq x < 9\\ 2.927 & 9 \leq x \end{cases}$$

Solution 8.16

$$\hat{S}_{0}(7) = \exp(-\hat{H}_{0}(7)) = 0.501$$
 (Solution 8.15)
 $\hat{K} = 0.618$ (Solution 8.14)
 $\Rightarrow \hat{S}_{0}(7)^{\hat{K}} = 0.652$

Solution 8.17

$$\operatorname{var}(\hat{\beta}_{1} z_{1} + \hat{\beta}_{2} z_{2}) = (z_{1})^{2} \operatorname{var}(\hat{\beta}_{1}) + (z_{2})^{2} \operatorname{var}(\hat{\beta}_{2}) + 2 z_{1} z_{2} \operatorname{cov}(\hat{\beta}_{1}, \hat{\beta}_{2})$$
$$= 0.00005(z_{1})^{2} + 0.00002(z_{2})^{2} + 0.00002 z_{1} z_{2}$$

Solution 8.18

The distribution of $\hat{\beta}_1 z_1 + \hat{\beta}_2 z_2$ is approximately normal with mean $\beta_1 z_1 + \beta_2 z_2$ and variance equal to the result in Solution 8.17.

Solution 8.19

From Solution 8.18, it follows that the distribution of $-\hat{\beta}_1 + 6\hat{\beta}_2$ is approximately normal with mean $-\beta_1 + 6\beta_2$ and variance:

$$E\left[-\hat{\beta}_1 + 6\hat{\beta}_2\right] = -\beta_1 + 6\beta_2 = 0.06200$$

$$\operatorname{var}\left(-\hat{\beta}_1 + 6\hat{\beta}_2\right) = 0.00005 \times (-1)^2 + 0.00002 \times (6)^2 + 0.00002 \times (-6) = 0.00065$$

The 90% confidence interval for $-\beta_1 + 6\beta_2$ is thus:

$$0.06200 \pm 1.645 \times \sqrt{0.00065}$$

This interval is: (0.020, 0.104).

The 95% confidence interval for $-\beta_1 + 6\beta_2$ (working exactly as in the previous question) is:

$$0.06200 \pm 1.96\sqrt{0.00065} = (0.01203, 0.11197)$$

The relative risk of a female who consumes 8 fluid ounces of alcohol per day (Life 1) to a male who consumes 2 fluid ounces per day (Life 2) is:

$$K_{12} = \frac{g(0,8)}{g(1,2)} = \frac{e^{8\beta_2}}{e^{\beta_1 + 2\beta_2}} = e^{-\beta_1 + 6\beta_2}$$

So we can obtain the 95% confidence interval for the relative risk by applying the exponential function to the endpoints of the interval just obtained:

$$\left(e^{0.0120}, e^{0.1120}\right) = \left(1.012, 1.118\right)$$