

Distribution Analysis

Variable: C1

Censoring Information	Count
Uncensored value	5
Right censored value	1

Nonparametric Estimates

Characteristics of Variable

Mean	Standard Error	95,0% Normal CI	
		lower	upper
494,1667	84,0765	329,3797	658,9537

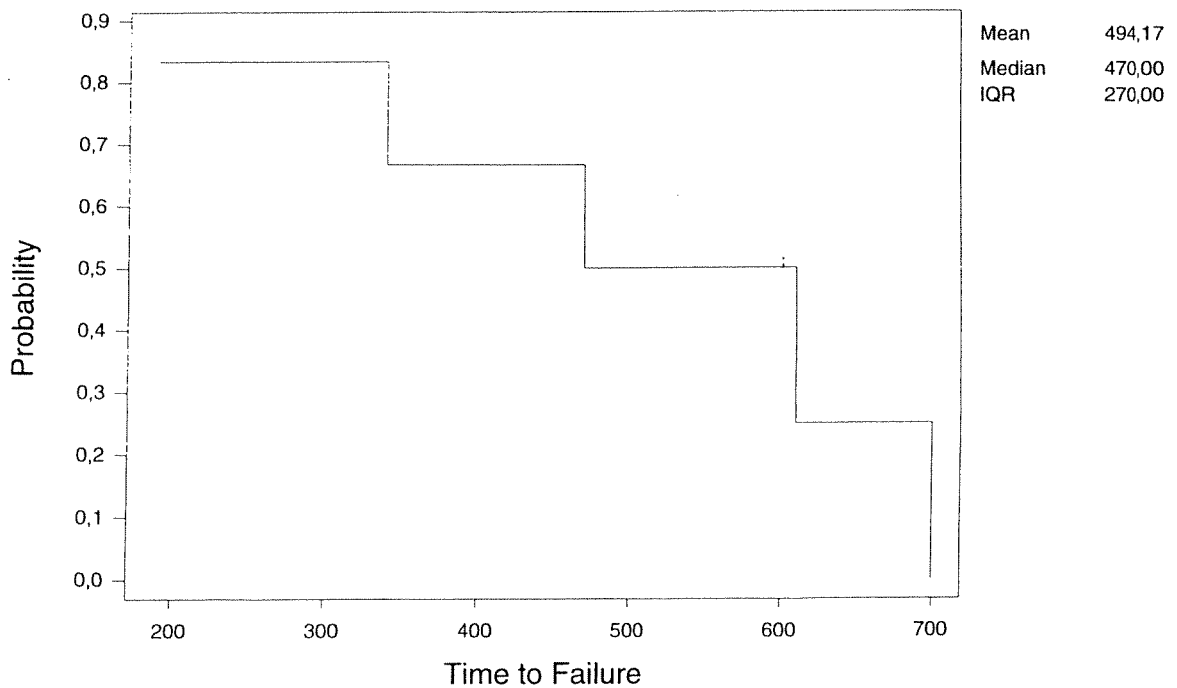
Median = 470,0000
 IQR = 270,0000 Q1 = 340,0000 Q3 = 610,0000

Kaplan-Meier Estimates

Time	Number at Risk	Number Failed	Survival Probability	Standard Error	95,0% Normal CI Lower	Upper
190,0000	6	1	0,8333	0,1521	0,5351	1,0000
340,0000	5	1	0,6667	0,1925	0,2895	1,0000
470,0000	4	1	0,5000	0,2041	0,0999	0,9001
610,0000	2	1	0,2500	0,2041	0,0000	0,6501
700,0000	1	1	0,0000	0,0000	0,0000	0,0000

Nonparametric Survival Plot for C1

Kaplan-Meier Method
 Censoring Column in C2



2a $d_i \begin{cases} 1 & T_i \leq C_i \\ 0 & T_i > C_i \end{cases} \quad Y_i = \min(T_i, C_i)$

$$l(Y_1, \dots, Y_n) = \prod_{i=1}^n f(Y_i)^{d_i} R(Y_i)^{1-d_i}$$

$$L = \ln(l) = \sum_{i=1}^n [d_i \ln f(Y_i) + (1-d_i) \ln R(Y_i)]$$

b $f(y_i) = \alpha d^{\alpha-1} y_i^{-\alpha} e^{-\lambda y_i} \quad R(y_i) = e^{-(\lambda y_i)^\alpha}$

$$L = \sum_{i=1}^n d_i [\ln \alpha + \alpha \ln d + (\alpha-1) \ln y_i] + \sum_{i=1}^n (-\lambda^\alpha y_i^\alpha)$$

$$\frac{\partial L}{\partial \lambda} = \sum_{i=1}^n \left[\frac{d_i \cdot \alpha}{\lambda} - \alpha \lambda^{\alpha-1} \cdot y_i^\alpha \right] = 0$$

$$\alpha \sum_{i=1}^n d_i - \alpha \lambda^\alpha \cdot \sum_{i=1}^n y_i^\alpha = 0$$

$$\lambda^\alpha = \frac{\sum_{i=1}^n d_i}{\sum_{i=1}^n y_i^\alpha} = \frac{d}{\sum_{i=1}^n y_i^\alpha} \quad \lambda^* = \left(\frac{d}{\sum_{i=1}^n y_i^\alpha} \right)^{\frac{1}{\alpha}}$$

$d = \sum d_i$

$d = 5 \quad \alpha = 3$

$$\lambda^* = \frac{5^{1/3}}{100 (1.9^3 + 3.4^3 + 4.7^3 + 5.4^3 + 6.1^3 + 7^3)^{1/3}}$$

$$\lambda^* = \frac{5^{1/3}}{100 (877.4)^{1/3}}$$

$\lambda^* \approx 0.001786$
 $\lambda^* \approx 0.0018$

$\lambda^* \approx 0.0018$

$$d) \frac{\partial^2 L}{\partial \lambda^2} = \alpha \cdot d \cdot \left(-\frac{1}{\lambda^2}\right) - \alpha(\alpha-1) \lambda^{\alpha-2} \sum_{i=1}^n Y_i^\alpha$$

$$-\frac{\partial^2 L}{\partial \lambda^2} = \frac{\alpha d}{\lambda^2} + \alpha(\alpha-1) \lambda^{\alpha-2} \sum_{i=1}^n Y_i^\alpha$$

$$\lambda^\alpha \sum_{i=1}^n Y_i^\alpha = d$$

$$-\frac{\partial^2 L}{\partial \lambda^2} \Big|_{\lambda=\lambda^*} = \frac{\alpha d}{\lambda^{*2}} + \alpha(\alpha-1) \frac{d}{\lambda^{*2}}$$

$$= \frac{\alpha^2 d}{\lambda^{*2}} = \frac{9 \cdot 5}{0.0018^2} = \underline{\underline{13,890.00}}$$

$$\underline{\underline{\text{Var}(\lambda^*)}} \approx \frac{0.0018}{45} = \underline{\underline{7.2 \cdot 10^{-8}}}$$

$$= \underline{\underline{0.00027^2}} \approx 0.0003^2$$

$$\lambda^* \sim N(\lambda, \text{Var}(\lambda^*)) \sim N(\lambda, 0.0003^2)$$

$\left. \begin{array}{l} \frac{\lambda^* - \lambda}{0.0003} \sim N \\ \lambda^* \pm 1.645 \cdot 0 \end{array} \right\}$

$$c) P(T > 600) \approx e^{-(0.0013 \cdot 600)^3}$$

$$e^{-1.26} = \underline{\underline{0.284}}$$

From Opps 1: $R(600) \approx \underline{\underline{0.500}}$

$$e) \quad H_0: \lambda = 0.002$$

$$L(\alpha, \lambda) = d \left[\ln \lambda^\alpha + \alpha \ln d + (\alpha - 1) \sum \delta_i \ln(Y_i) \right. \\ \left. - \lambda^\alpha \sum_{i=1}^n Y_i^\alpha \right]$$

$$W = 2 \left[L(3, \lambda^*) - L(3, 0.002) \right] \quad \sim \chi^2_{\text{under } H_0}$$

$$2 \left\{ d \alpha \left[\ln(0.0018) - \ln(0.002) \right] \right. \\ \left. - \sum_{i=1}^n Y_i^\alpha \left[0.0018^\alpha - 0.002^\alpha \right] \right\}$$

$$\sum_{i=1}^n Y_i^\alpha = \frac{d}{(\lambda^*)^\alpha} \quad d = 5, \quad \alpha = 3$$

$$\underline{W} = 2 \left\{ 5 \cdot 3 \cdot \ln\left(\frac{18}{20}\right) - \frac{5}{0.0018^3} \left[0.0018^3 - 0.002^3 \right] \right\}$$

$$= 2 \left[-1.58 + 5 \left[-1 + \left(\frac{20}{18}\right)^3 \right] \right]$$

$$= \underline{0.56} \quad \text{Illus. for } H_0. \quad 1.85$$

0 MS 3

Overlevelsesfunksjon

A 8 P s. 66
H 2 R s. 428
423

$$a) R(t/\underline{s}) = R_0(\lambda(\underline{s}) \cdot t)$$

$$\text{der funksjon } \lambda(\underline{s}) = \begin{cases} C \cdot s^a \\ C e^{-b/s} \\ C s e^{-b/s} \end{cases}$$

$$\begin{aligned} E(T|\underline{s}) &= \int_0^{\infty} R(t|\underline{s}) dt = \int_0^{\infty} R_0(\lambda(\underline{s}) \cdot t) dt \\ &= \int_0^{\infty} R_0(u) \frac{du}{\lambda(\underline{s})} = \frac{E(T|\underline{s}=0)}{\lambda(\underline{s})} \end{aligned}$$

b)

$$\lambda(\underline{s}) = e^{\beta_1 s_1 + \dots + \beta_n s_n}$$

Weibull regresjon:

$$(2) R(t/\underline{s}) = e^{-[\lambda(\underline{s}, \underline{\beta}) \cdot t]^\alpha}$$

$$\text{der } \lambda(\underline{s}, \underline{\beta}) = e^{\beta_1 s_1 + \dots + \beta_n s_n}$$

Altså, skalaparameter, λ , avhenger av stressorer s_1, \dots, s_n

(2) spesialtilfelle av (1).

0.119 4

a posteriori $f(\lambda/x) \propto \frac{(\lambda t)^x}{x!} e^{-\lambda t} \cdot \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\beta \lambda}$

$\propto \lambda^{x+\alpha-1} e^{-(t+\beta)\lambda}$ Gamma(x+alpha, t+beta)

b) $\hat{\lambda} = \frac{x+\alpha}{t+\beta}$ H&K S. 446

a priori $\frac{\alpha}{\beta} = 0.1$

$\hat{\beta} = 20$ or $\alpha = 2$

c) $\hat{\lambda} = \frac{12 + 2}{134 + 20} = \frac{14}{154} = 0.091$

klammert: $\frac{12}{134} = 0.09$ H&K S. 446

$P \left[\frac{z_{1-\epsilon/2, 2(\alpha+x)}}{2(\beta+t)} < \lambda < \frac{z_{\epsilon/2, 2(\alpha+x)}}{2(\beta+t)} \right] = 1 - \epsilon$

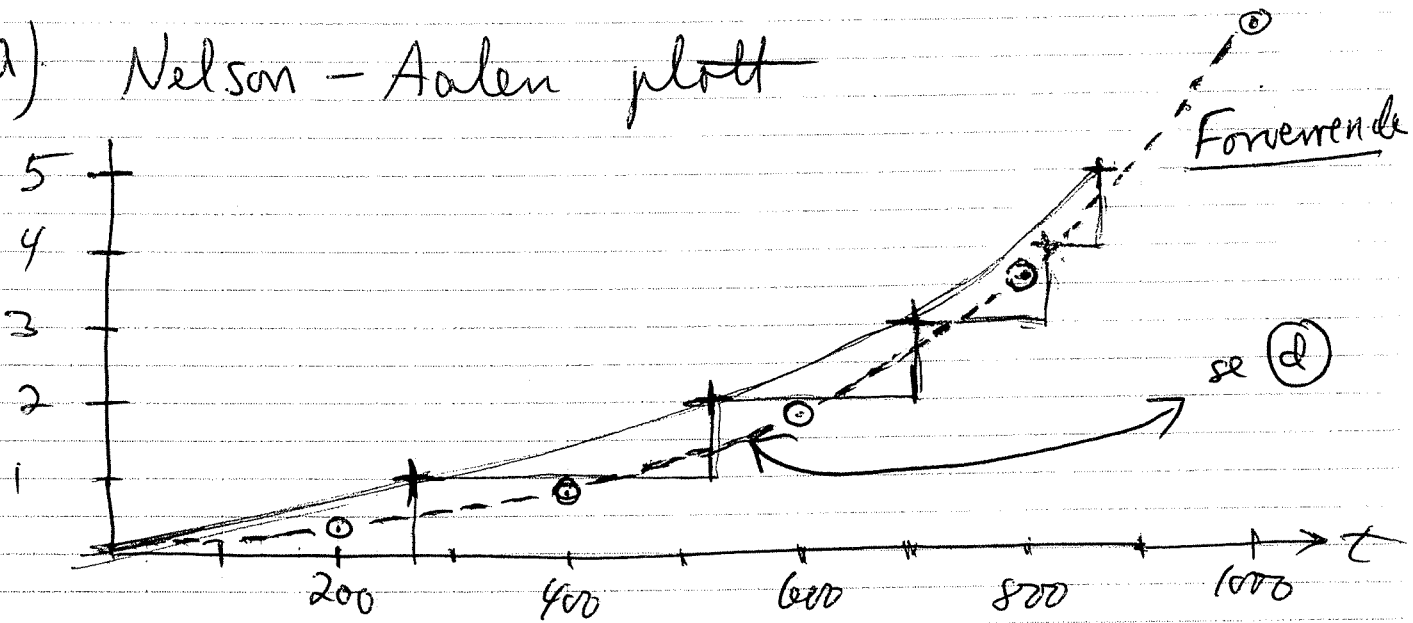
$P \left(\frac{z_{1-\epsilon/2, 2(\alpha+x)}}{2(\beta+t)} < \lambda < \frac{z_{\epsilon/2, 2(\alpha+x)}}{2(\beta+t)} \right) = 1 - \epsilon$

$\alpha = 2$
 $x = 12$
 $\beta = 20$
 $t = 134$

$\left(\frac{z_{0.95; 28}}{2 \cdot 154} < \lambda < \frac{z_{0.05; 28}}{2 \cdot 154} \right)$

Oppg 5

a) Nelson - Aalen plott



$$\bar{S} = \frac{1}{n} \sum_{j=1}^n S_j \sim N\left(\tau/2, \left(\tau/\sqrt{12n}\right)^2\right)$$

$$U = \frac{\bar{S} - \tau/2}{\tau/\sqrt{12n}} \sim N(0,1)$$

$U > 0$ forværende

$U < 0$ forbedrende

$$\text{Før } U = \frac{632 - 450}{900/\sqrt{100}} = \underline{1,57}$$

H_0 : ingen trend

Blev ikke forkastet ved nivå 5%

b) log-linear model $\omega(t) = e^{\alpha + \beta t}$

$$\underline{W(t)} = \frac{1}{\beta} [e^{\alpha + \beta t} - e^{\alpha}] = \frac{e^{\alpha}}{\beta} [e^{\beta t} - 1]$$

$\beta=0, \omega(t) = e^{\alpha}$ $\underline{W(t)} = e^{\alpha} \cdot t$

c) $l = \omega(s_1) \dots \omega(s_n) \cdot e^{-\int_0^{\tau} \omega(t) dt}$

$$= e^{n\alpha + \beta \sum_{i=1}^n s_i} \cdot e^{-e^{\alpha} [e^{\beta \tau} - 1] / \beta}$$

$$L = \ln(l) = n\alpha + \beta \cdot \sum_{i=1}^n s_i - e^{\alpha} [e^{\beta \tau} - 1] / \beta$$

$$(1) \frac{\partial L}{\partial \beta} = \sum_{i=1}^n s_i - e^{\alpha} \left\{ \frac{1}{\beta} \cdot \tau e^{\beta \tau} - \frac{1}{\beta^2} (e^{\beta \tau} - 1) \right\}$$

$$(2) \frac{\partial L}{\partial \alpha} = n - e^{\alpha} [e^{\beta \tau} - 1] / \beta$$

$$(2) \rightarrow (2') \quad \underbrace{e^{\alpha^*} = \frac{n/\beta^*}{e^{\beta^* \tau} - 1}}_{\text{inset (1)}}$$

$$0 = \sum_{i=1}^n s_i - \frac{n/\beta^*}{e^{\beta^* \tau} - 1} \left\{ \frac{1}{\beta^*} \tau e^{\beta^* \tau} - \frac{1}{\beta^{*2}} (e^{\beta^* \tau} - 1) \right\}$$

$$0 = \sum_{i=1}^n s_i - \frac{n \tau e^{\beta^* \tau}}{e^{\beta^* \tau} - 1} + \frac{n}{\beta^*} \quad \underline{n = N(\tau)}$$

$$(1') \quad \underbrace{\sum_{i=1}^n s_i + \frac{N(\tau)}{\beta^*}}_{\text{Cur } \beta^*} = \frac{N(\tau) \cdot \tau}{1 - e^{-\beta^* \tau}}$$

$$(2') \quad \alpha^* = \ln \left[\frac{N(\tau) \beta^*}{e^{\beta^* \tau} - 1} \right]$$

d) Setzt $\sum s_i = 3160$ $n = N(\tau) = 5$ $\tau = 900$

Als (1') findet $\underline{\beta^*} = 0.00301 \approx \underline{0.003}$
(numerisch)

$$\alpha^* = \ln \left[\frac{5 \cdot 0.003}{e^{0.003 \cdot 900} - 1} \right] \approx \underline{-6.83}$$

$$\hat{W}(t) = \frac{e^{-6.83}}{0.003} \left[e^{0.003 \cdot t} - 1 \right] = 0.36 e^{0.003 \cdot t}$$

t	$\hat{W}(t)$
$t = 200$	0.3
$t = 400$	0.84
$t = 600$	1.8
$t = 800$	3.6
$t = 1000$	6.9

Bra übereinstimmend