

HW8

Zhuodiao Kuang

2023-11-18

Problem 1

Since $T_i \sim LN(\mu, \sigma^2)$, we have the p.d.f. of the log-normal:

$$f(t) = \frac{1}{t\sigma\sqrt{2\pi}} \exp\left(-\frac{(\ln t - \mu)^2}{2\sigma^2}\right)$$

The log-likelihood of the sample $\{t_1, t_2, \dots, t_n\}$ is:

$$LL = -\frac{n}{2} \ln(2\pi) - \frac{n}{2} \ln \sigma^2 - \frac{1}{2\sigma^2} \sum_{i=1}^n (\ln t_i - \mu)^2 - \sum_{i=1}^n \ln t_i$$

Taking the F.O.C. with respect to μ and σ , and set them to zero, we can get:

$$\begin{aligned} \frac{\partial LL}{\partial \mu} &= \frac{1}{\sigma^2} \left(\sum_{i=1}^n \ln t_i - \mu \right) = 0 \\ \frac{\partial LL}{\partial \sigma^2} &= -\frac{n}{2\sigma^2} + \frac{\sum_{i=1}^n (\ln t_i - \mu)^2}{2\sigma^4} = 0 \end{aligned}$$

We can solve them MLE $\hat{\mu}, \hat{\sigma}^2$ from the above equations:

$$\begin{aligned} \hat{\mu} &= \frac{1}{n} \sum_{i=1}^n \ln t_i \\ \hat{\sigma}^2 &= \frac{1}{n} \sum_{i=1}^n (\ln t_i - \hat{\mu})^2 \end{aligned}$$

Therefore, MLEs have closed form.

Problem 2

```
##
## Call:
## survreg(formula = Surv(futime, fustat) ~ ecog.ps, data = ovarian,
##         dist = "exponential")
##               Value Std. Error      z      p
## (Intercept)   7.893      0.971   8.13 4.3e-16
## ecog.ps       -0.475      0.586  -0.81   0.42
##
```

```
## Scale fixed at 1
##
## Exponential distribution
## Loglik(model)= -97.7   Loglik(intercept only)= -98
##   Chisq= 0.67 on 1 degrees of freedom, p= 0.41
## Number of Newton-Raphson Iterations: 4
## n= 26

##
## Call:
## survreg(formula = Surv(futime, fustat) ~ ecog.ps, data = ovarian,
##         dist = "weibull")
##               Value Std. Error      z      p
## (Intercept)  7.766      0.915  8.49 <2e-16
## ecog.ps      -0.431      0.534 -0.81  0.42
## Log(scale)   -0.108      0.254 -0.43  0.67
##
## Scale= 0.897
##
## Weibull distribution
## Loglik(model)= -97.6   Loglik(intercept only)= -98
##   Chisq= 0.69 on 1 degrees of freedom, p= 0.41
## Number of Newton-Raphson Iterations: 5
## n= 26
```

Because both tests of two distributions generate relatively high p-values, so the survival data, by ecog status, doesn't follow the exponential distribution or the Weibull distribution.

Problem 3

The likelihood of Weibull distribution is:

$$\begin{aligned}
 L(\beta) &= \prod_{i=1}^n h(T_i | Z_i)^{\Delta_i} S(T_i) \\
 &= \prod_{i=1}^n (h_0(T_i) e^{\beta Z_i})^{\Delta_i} e^{-\lambda T_i^\alpha} \\
 &= \prod_{i=1}^n (\lambda \alpha T_i^{\alpha-1} e^{\beta Z_i})^{\Delta_i} e^{-\lambda T_i^\alpha} \\
 &= (\lambda \alpha \times 16^{\alpha-1} e^{\beta})^1 e^{-\lambda \times 16^\alpha} \times e^{-\lambda \times 20^\alpha} \\
 &\times (\lambda \alpha \times 12^{\alpha-1})^1 e^{-\lambda \times 12^\alpha} \times e^{-\lambda \times 14^\alpha} \\
 &\times (\lambda \alpha \times 11^{\alpha-1})^1 e^{-\lambda \times 11^\alpha} \\
 &\times (\lambda \alpha \times 9^{\alpha-1} e^{\beta})^1 e^{-\lambda \times 9^\alpha} \\
 &= (\lambda \alpha)^4 (16^{\alpha-1} 12^{\alpha-1} 11^{\alpha-1} 9^{\alpha-1}) e^{2\beta - \lambda(16^\alpha + 20^\alpha + 12^\alpha + 14^\alpha + 11^\alpha + 9^\alpha)} \\
 &= (\lambda \alpha)^4 19008^{\alpha-1} e^{2\beta - \lambda(16^\alpha + 20^\alpha + 12^\alpha + 14^\alpha + 11^\alpha + 9^\alpha)}
 \end{aligned}$$

Problem 4

```
##
```

```
## Call:
## survreg(formula = Surv(time, event == 1) ~ trt, data = leu_dat,
##       dist = "loglogistic")
##               Value Std. Error      z      p
## (Intercept)  1.893      0.208  9.12 < 2e-16
## trt6-MP      1.265      0.326  3.89  1e-04
## Log(scale)  -0.604      0.150 -4.02 5.7e-05
##
## Scale= 0.547
##
## Log logistic distribution
## Loglik(model)= -107.7   Loglik(intercept only)= -115.4
##  Chisq= 15.38 on 1 degrees of freedom, p= 8.8e-05
## Number of Newton-Raphson Iterations: 4
## n= 42
```

From the AFT model summary, we can see that the coefficient corresponds to the treatment effect is significant, which means that the treatment 6-MP has a significant effect on the survival of patients with Acute Myelogenous Leukemia.

The corresponding model is:

$$\log T_i = 1.8927 + 1.2655 \times I(\text{trt}_i = \text{6-MP})$$