





$$\frac{50(100)^3}{2,000,000} = 25 \text{ seconds}$$


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5. (20 pts) Given a sample of size  $n$  from the lognormal distribution

$$f(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma x}} e^{-(\log x - \mu)^2 / 2\sigma^2}, \quad x > 0, \mu \in \mathcal{R}, \sigma > 0,$$

find the gradient and Hessian matrices to be used in minimizing via Newton–Raphson the negative of the log likelihood of  $\mu$  and  $\sigma^2$ .

**SOLUTION:**

(Note: The original question had the minus sign in the exponent inside the parentheses instead of outside. This solution assumes the correct expression for  $f$ .)

We have

$$\begin{aligned} L(\theta) &= L\left(\begin{pmatrix} \mu \\ \sigma^2 \end{pmatrix}\right) = -\sum_{i=1}^n \log f(x_i) \\ &= -\sum_{i=1}^n \left[ -\frac{1}{2} \log 2\pi - \frac{1}{2} \log \sigma^2 - \log x_i - (\log x_i - \mu)^2 / 2\sigma^2 \right] \\ &= \frac{n}{2} \log 2\pi + \frac{n}{2} \log \sigma^2 + \sum_{i=1}^n \log x_i + \frac{1}{2\sigma^2} \sum_{i=1}^n (\log x_i - \mu)^2, \end{aligned}$$

and so

$$g = \begin{pmatrix} \frac{\delta L}{\delta \mu} \\ \frac{\delta L}{\delta \sigma^2} \end{pmatrix} = \begin{pmatrix} -\frac{1}{\sigma^2} \sum_{i=1}^n (\log x_i - \mu) \\ \frac{n}{2\sigma^2} - \frac{1}{2\sigma^4} \sum_{i=1}^n (\log x_i - \mu)^2 \end{pmatrix}$$

and

$$H = \begin{bmatrix} \frac{\delta^2 L}{\delta \mu^2} & \frac{\delta^2 L}{\delta \mu \delta \sigma^2} \\ \text{sym} & \frac{\delta^2 L}{\delta (\sigma^2)^2} \end{bmatrix} = \begin{bmatrix} \frac{n\mu}{\sigma^2} & \frac{1}{\sigma^4} \sum_{i=1}^n (\log x_i - \mu) \\ \text{sym} & -\frac{n}{2\sigma^4} + \frac{1}{\sigma^6} \sum_{i=1}^n (\log x_i - \mu)^2 \end{bmatrix}.$$


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6. (20 pts) How would you use the U(0,1) distribution in the rejection method to generate random numbers from the beta distribution having pdf

$$f(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}, \quad x \in (0, 1), \alpha > 0, \beta > 0,$$

for  $\alpha = \beta$  for integer values of  $\alpha > 1$ ? For a given value of  $\alpha$ , what is the probability of rejection? What is this probability for  $\alpha = 3$ ?

**SOLUTION:**

To use the rejection method we must find a constant  $c$  satisfying

$$c = \max_{x \in (0,1)} \frac{f(x)}{g(x)},$$

or, since  $g(x) = 1$  over  $x \in (0, 1)$ , we must find the maximum value of  $f(x)$ . But we can use the first and second derivatives of  $f$  to show that the max of  $f$  occurs at  $x = 1/2$  and so

$$c = \frac{\Gamma(2\alpha)}{\Gamma^2(\alpha)4^{\alpha-1}} = \frac{(2\alpha - 1)!}{[(\alpha - 1)!]^2 4^{\alpha-1}},$$

since  $\alpha$  is assumed to be an integer. For  $\alpha = 3$ ,  $c = 15/8$ , which gives a rejection probability of  $1 - 1/c = 7/15$ . The rejection method is then

1. Generate  $U_1$  from  $U(0,1)$
2. Generate  $Z = cU_2$ , where  $U_2$  is another  $U(0,1)$  independent of  $U_1$ .
3. If  $Z < f(U_1)$ , return  $X = U_1$ , else return to (1).

7. (20 pts) Write an Splus function `ptnorm(sig1,sig2)` that will superimpose plots of the  $N(0,\sigma^2)$  pdf for 101 equally spaced values of  $\sigma$  in `[sig1,sig2]`, with each pdf plotted at 301 equally spaced points from the .001 to .999 quantile of the distribution. The function must call `plot` to set up the axes and then do a `for` loop that will call `lines` to draw each pdf.

**SOLUTION:**

The following version of `ptnorm` is slightly more general as it has arguments corresponding to number of pdf's to draw and the number of points to use. The important thing in the function is setting up the axes using `plot`. We know the pdf with the largest variance is the most spread out while the one with the smallest variance is the tallest. These facts are used in the `plot` call below. The graph on the last page of this solution set is obtained by `ptnorm(1,2,11,101)`.

```
ptnorm <- function(sig1=1,sig2=2,nplts=101,nxs=301)
{
  plot(c(qnorm(.001,0,sig2),qnorm(.999,0,sig2)),c(0,dnorm(0,0,sig1)),
       type='n',xlab='x',ylab='f(x)',
       main=paste(nplts, ' N(0,sigma^2) pdfs for sigma in [',
       round(sig1,2),',',round(sig2,2),']'))
  for(i in 1:nplts)
  {
    sig <- sig1+(i-1)*(sig2-sig1)/(nplts-1)
    lines(x <- seq(qnorm(.001,0,sig),qnorm(.999,0,sig),length=nxs),
          dnorm(x,0,sig))
  }
}
```

11  $N(0, \sigma^2)$  pdfs for  $\sigma$  in [ 1 , 2 ]

