



APL747

Homework - 2

Q.1) $f_z(z) = \alpha \beta z^{\beta-1} e^{-\alpha z^\beta}; z \geq 0$

For $z < 0$: $F_z(z) = P(Z < z) = 0$

For $z \geq 0$: $F_z(z) = \int_0^z \alpha \beta z^{\beta-1} e^{-\alpha z^\beta} dz$

Let $\alpha \cdot t = \alpha z^\beta \Rightarrow dt = \alpha \beta z^{\beta-1} dz$; $t = \alpha z^\beta \Rightarrow z = (t/\alpha)^{1/\beta} = t^{1/\beta}$

$$\therefore F_z(z) = \int_0^t e^{-t} dt = -e^{-\alpha z^\beta} \Big|_0^z = 1 - e^{-\alpha z^\beta}$$

Now, $y = 1 - e^{-\alpha z^\beta}$

for inverse (F^{-1}): $z = 1 - e^{-\alpha y^\beta} \Rightarrow e^{-\alpha y^\beta} = 1 - z \Rightarrow -\alpha y^\beta = \ln(1 - z)$

$$\therefore y = \left[\frac{\ln \frac{1}{1-z}}{\alpha} \right]^{1/\beta} \Rightarrow F^{-1}(u^{(i)}) = \left[\frac{\ln \frac{1}{(1-u^{(i)})^\alpha}}{\alpha} \right]^{1/\beta} = \left[\frac{-1}{\alpha} \ln(1 - u^{(i)}) \right]^{1/\beta}$$

$$\therefore x^{(i)} \sim F^{-1}(u^{(i)})$$

2) We have already been given the cdf of the distribution.

$$F_z(z) = e^{-e^{-\alpha(z-\mu)}}; -\infty < z < \infty$$

Now,

$$z = e^{-e^{-\alpha(y-\mu)}} \Rightarrow \ln z = -e^{-\alpha(y-\mu)} \Rightarrow \ln(-\ln z) = -\alpha(y-\mu)$$

$$\Rightarrow y = \mu - \frac{1}{\alpha} \ln(-\ln z)$$

$$\therefore F^{-1}(u^{(i)}) = \mu - \frac{1}{\alpha} \ln(-\ln u^{(i)})$$

$$\therefore x^{(i)} \sim F^{-1}(u^{(i)})$$



$$3) f_z(z) = \frac{1}{B(p,q)} z^{p-1} (1-z)^{q-1}; \quad 0 \leq z \leq 1$$

Now,

$$F(z) = \int_0^z \frac{1}{B(p,q)} z^{p-1} (1-z)^{q-1} dz$$

∴ The exact analytical solution of the above integral is very difficult to compute.

$$\therefore u^{(i)} B(p,q) = \int_0^z z^{p-1} (1-z)^{q-1} dz$$

We ~~use numerical~~

CDF

Thus, we use the inbuilt inverse [^]function of the incomplete Beta distribution.

Q.6. According to the perturbation theory,

~~Method~~ ①: Laplace approximation of the vector $\underline{p} \underline{u}$

$$\underline{u} \underline{p} = [c \ k \ \alpha]^T$$

Since the three parameters are independent, their mutual covariance is 0.

For a logNormal distribution $(\mu, \sigma) = e^{\mu - \sigma^2} = \underline{u}_0$

Taking Taylor series expansion of $\log(p(\underline{u}_i))$ about \underline{u}_0 :

$$\log(p(u)) = \log(p(u_0)) + \underbrace{\frac{d \log(p(u))}{du}}_{u_0 \text{ is mode}} \bigg|_{u_0} (u - u_0) + \frac{1}{2} \frac{d^2 \log(p(u))}{du^2} \bigg|_{u_0} (u - u_0)^2$$

$$\therefore \log(p(u)) = \log(p(u_0)) + \frac{1}{2} \frac{d^2 \log(p(u))}{du^2} \bigg|_{u_0}$$



$$\therefore p(u) \propto \exp \left[\frac{d^2 \log p(u)}{du^2} \bigg|_{u_0} \frac{(u-u_0)^2}{2} \right]$$

For logNormal,

$$p(u) = \frac{1}{u \sqrt{2\pi\sigma^2}} \exp \left[-\frac{(\ln u - \mu)^2}{2\sigma^2} \right]$$

$$\therefore \log(p(u)) = -\log u - \frac{1}{2} \log(2\pi\sigma^2) - \frac{(\ln u - \mu)^2}{2\sigma^2}$$

$$\frac{d^2 \log(p(u))}{du^2} = \left[-\frac{1}{u} - \frac{(\ln u - \mu)}{\sigma^2} \left(\frac{1}{u} \right) \right] = -\frac{1}{u} \left[1 + \frac{\ln u - \mu}{\sigma^2} \right]$$

$$\therefore \frac{(u-u_0)^2}{2} \frac{d^2 \log(p(u))}{du^2} = \frac{-1}{2}$$

$$\begin{aligned} \frac{d^2 \log(p(u))}{du^2} \bigg|_{u_0} &= \frac{-1}{u^2} \left(1 + \frac{\ln u - \mu}{\sigma^2} \right) \frac{1}{u^2} \left(1 + \frac{\ln u - \mu}{\sigma^2} \right) - \frac{1}{u} \left[\frac{1}{u\sigma^2} \right] \\ &= \frac{1}{u_0^2} \left[1 + \frac{\ln u_0 - \mu}{\sigma^2} - \frac{1}{\sigma^2} \right] \end{aligned}$$

$$\begin{aligned} \text{for } u_0 = e^{\mu - \sigma^2}: \quad \frac{d^2 \log p(u)}{du^2} \bigg|_{u_0} &= \frac{1}{e^{2(\mu - \sigma^2)}} \left[\sigma^2 - 1 - \mu + \mu - \sigma^2 \right] \\ &= \frac{-1}{e^{2(\mu - \sigma^2)} \sigma^2} \end{aligned}$$

$$\therefore \text{Var}[u] = - \left(\frac{-1}{e^{2(\mu - \sigma^2)} \sigma^2} \right)^{-1} = \sigma^2 e^{2(\mu - \sigma^2)}$$

$$\therefore p(\underline{u}) = \mathcal{N} \left(\underline{u}_0, \underline{\Sigma} \right)$$

$$\text{where, } \underline{u}_0 = \begin{bmatrix} e^{\mu_c - \sigma_c^2} & e^{\mu_k - \sigma_k^2} & e^{\mu_x - \sigma_x^2} \end{bmatrix}^T$$

$$\underline{\Sigma} = \begin{bmatrix} \sigma_c^2 e^{2(\mu_c - \sigma_c^2)} & 0 & 0 \\ 0 & \sigma_k^2 e^{2(\mu_k - \sigma_k^2)} & 0 \\ 0 & 0 & \sigma_x^2 e^{2(\mu_x - \sigma_x^2)} \end{bmatrix}$$



For the given system:

$$\text{Let } y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

$$\therefore \frac{d}{dt} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} -cy_2 - ky_1 - \alpha y_1^3 + A_1 \sin(\pi t) + A_2 \sin(2\pi t) \end{bmatrix}$$

Initial Cond's
Now, $y_1(0) = 0.1$; $y_2(0) = 0$

$$\frac{d}{dt} \left(\frac{\partial y_1}{\partial y_j} \right) = \frac{\partial}{\partial y_j} \left(\frac{dy_1}{dt} \right) = \frac{\partial (y_2)}{\partial y_j} = \frac{\partial y_2}{\partial y_j}$$

$$\Rightarrow \frac{d}{dt} \left(\frac{\partial y_1}{\partial y_j} \right) = \frac{\partial y_2}{\partial y_j} \quad \text{--- 3 eqns: } j = 1, 2, 3$$

$$\begin{aligned} \frac{d}{dt} \left(\frac{\partial y_2}{\partial c} \right) &= \frac{\partial}{\partial c} \left(\frac{dy_2}{dt} \right) = \frac{\partial}{\partial c} (-cy_2 - ky_1 - \alpha y_1^3 + A_1 A_2 \sin(\pi t) \sin(2\pi t)) \\ &= -y_2 - c \frac{\partial y_2}{\partial c} - k \frac{\partial y_1}{\partial c} - 3\alpha y_1^2 \frac{\partial y_1}{\partial c} \end{aligned}$$

$$\text{dy. } \frac{d}{dt} \left(\frac{\partial y_2}{\partial k} \right) = -c \frac{\partial y_2}{\partial k} - y_1 - k \frac{\partial y_1}{\partial k} - 3\alpha y_1^2 \frac{\partial y_1}{\partial k}$$

$$\frac{d}{dt} \left(\frac{\partial y_2}{\partial \alpha} \right) = -c \frac{\partial y_2}{\partial \alpha} - k \frac{\partial y_1}{\partial \alpha} - y_1^3 - 3\alpha y_1^2 \frac{\partial y_1}{\partial \alpha}$$

For initial value: $\frac{d}{dt} \left(\frac{\partial y_1(0)}{\partial y_j} \right) = \frac{\partial}{\partial y_j} \left(\frac{dy_1(0)}{dt} \right) = 0 \quad j = 1, 2, 3$