

STAT GU4261/GR5261 STATISTICAL METHODS IN FINANCE

SPRING 2018

HOMEWORK 6 SUGGESTED SOLUTION

DUE DATE: 11 MAR 2017 (SUN)

Note: In this suggested solution, R is used to compute the quantiles of normal and t -distributions if required. If you have used the rounded-off figures provided in the normal table, you will have slightly different results, but it is fine.

1. Denote μ and σ as the mean and standard deviation of the loss L , then $\mu = -0.002$ and $\sigma = 0.016$. The required probability is

$$P(L > \mu + \sigma) = P\left(\frac{L - \mu}{\sigma} \sim N(0, 1) > 1\right) = 0.159.$$

2. (a) Recall $-\text{VaR}(\alpha)$ is the α -th quantile of R , $-z_\alpha$ is the α -th quantile of $N(0, 1)$, hence $\text{VaR}(0.1) = -1000000[0.002 + 0.016(-z_{0.1})] = 18500$.

$$(b) \text{VaR}(0.05) = -1000[0.1 + 0.2(-z_{0.05})] = 229.$$

3. (a) If $-t_\alpha(\nu)$ denotes the α -th quantile of a t -distribution with degree of freedom ν , then $\text{VaR}(0.1) = -1000[0.002 + 0.016(-t_{0.1}(2))] = 28.2$.

$$(b) \text{VaR}(0.1) = -1000[0.002 + 0.016(-t_{0.1}(5))] = 21.6.$$

4. (a) Note that $w_A = 1/3$, $w_B = 2/3$, and $R_P = w_A R_A + w_B R_B \sim N(\mu_P, \sigma_P^2)$, where $\mu_P = w_A \mu_A + w_B \mu_B$ and $\sigma_P^2 = w_A^2 \sigma_A^2 + w_B^2 \sigma_B^2 + 2\rho w_A w_B \sigma_A \sigma_B$. If the returns are independent, then $\rho = 0$ and $\text{VaR}(0.05) = -1500[\mu_P + \sigma_P(-z_{0.05})] = 34.3$.

$$(b) \text{If } \rho = 0.3, \text{ then } \text{VaR}(0.05) = -1500[\mu_P + \sigma_P(-z_{0.05})] = 38.7.$$

$$(c) \text{If } \rho = -0.3, \text{ then } \text{VaR}(0.05) = -1500[\mu_P + \sigma_P(-z_{0.05})] = 29.4.$$

5. (a) We first compute the cdf, as it is also required in (b). When $a \leq -1$, we have

$$\begin{aligned} ZF(a) &= - \int_{-\infty}^a \frac{x+1}{(x^2+1)^2} dx = - \int_{-\infty}^a \frac{1/2}{(x^2+1)^2} d(x^2) - \int_{-\pi/2}^{\tan^{-1} a} \frac{\sec^2 x}{(\tan^2 x + 1)^2} dx \\ &= \frac{1}{2(x^2+1)} \Big|_{-\infty}^a - \int_{-\pi/2}^{\tan^{-1} a} \cos^2 x dx = \frac{1}{2(a^2+1)} - \int_{-\pi/2}^{\tan^{-1} a} \frac{1 + \cos 2x}{2} dx \\ &= \frac{1}{2(a^2+1)} - \left[\frac{x}{2} + \frac{\sin 2x}{4} \right]_{-\pi/2}^{\tan^{-1} a} = \frac{1}{2(a^2+1)} - \frac{\tan^{-1} a}{2} - \frac{\pi}{4} - \frac{\sin 2(\tan^{-1} a)}{4} \\ &= \frac{1}{2(a^2+1)} - \frac{\tan^{-1} a}{2} - \frac{\pi}{4} - \frac{\sin(\tan^{-1} a) \cos(\tan^{-1} a)}{2} \\ &= \frac{1}{2(a^2+1)} - \frac{\tan^{-1} a}{2} - \frac{\pi}{4} - \frac{a \cdot 1}{2\sqrt{a^2+1}\sqrt{a^2+1}} = -\frac{a-1}{2(a^2+1)} - \frac{\tan^{-1} a}{2} - \frac{\pi}{4}. \end{aligned}$$

This implies $ZF(-1) = \frac{1}{2} - \frac{-\pi/4}{2} - \frac{\pi}{4} = \frac{1}{2} - \frac{\pi}{8}$. When $a > -1$, we have

$$\begin{aligned}
ZF(a) &= ZF(-1) + \int_{-1}^a \frac{x+1}{(x^2+1)^2} dx = \frac{1}{2} - \frac{\pi}{8} + \int_{-1}^a \frac{x+1}{(x^2+1)^2} dx \\
&= \frac{1}{2} - \frac{\pi}{8} - \frac{1}{2(x^2+1)} \Big|_{-1}^a + \left[\frac{x}{2} + \frac{\sin 2x}{4} \right]_{-\pi/4}^{\tan^{-1} a} \\
&= \frac{1}{2} - \frac{\pi}{8} - \frac{1}{2(a^2+1)} + \frac{1}{4} + \frac{\tan^{-1} a}{2} + \frac{\pi}{8} + \frac{\sin 2(\tan^{-1} a)}{4} + \frac{1}{4} \\
&= 1 - \frac{1}{2(a^2+1)} + \frac{\tan^{-1} a}{2} + \frac{\sin 2(\tan^{-1} a)}{4} = 1 + \frac{a-1}{2(a^2+1)} + \frac{\tan^{-1} a}{2}.
\end{aligned}$$

It follows from $F(\infty) = 1$ that $Z = 1 - 0 + \frac{\pi/2}{2} = 1 + \pi/4$.

- (b) $\text{VaR}(0.05)$ is the value of a such that $F(a) = 0.95$. But $F(-1) = \frac{1/2 - \pi/8}{1 + \pi/4} < 0.95$, so we have $a > -1$. To this end, we equate the formula of $F(a)$ for $a > -1$ obtained above to 0.95 and solve for a (using equation solver, eg. **fzero** in MATLAB or **uniroot** in R) to get $a = 2.4637$.

6. (a) Recall $\text{VaR}(\alpha)$ is the $(1-\alpha)$ -th quantile of L , and $z_\alpha = -z_{1-\alpha}$ is the $(1-\alpha)$ -th quantile of $N(0, 1)$. If $L \sim N(\mu, \sigma^2)$, then $\text{VaR}(\alpha) = \mu + \sigma z_\alpha$, therefore

$$\begin{aligned}
\text{ES}(\alpha) &= E(L|L \geq \text{VaR}(\alpha)) = \frac{1}{\alpha} \int_{\text{VaR}(\alpha)}^{\infty} \frac{x}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\
&= \frac{1}{\alpha} \int_{\frac{\text{VaR}(\alpha)-\mu}{\sigma}}^{\infty} \frac{\mu + \sigma y}{\sqrt{2\pi\sigma^2}} e^{-\frac{y^2}{2}} \sigma dy = \frac{1}{\alpha} \left[\mu \int_{z_\alpha}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy + \sigma \int_{z_\alpha}^{\infty} \frac{y}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy \right] \\
&= \mu + \frac{\sigma}{\alpha} \left[-\frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \right]_{z_\alpha}^{\infty} = \mu + \frac{\sigma}{\alpha} \frac{1}{\sqrt{2\pi}} e^{-\frac{z_\alpha^2}{2}} = \mu + \frac{\sigma}{\alpha} \phi(z_\alpha).
\end{aligned}$$

- (b) Using the above formula, we have $\text{ES}(0.05) = 100000(-0.04 + \frac{0.18}{0.05} \phi(z_\alpha)) = 33129$.

- (c) We calculate $\mu_P = w_A \mu_A + w_B \mu_B$ and $\sigma_P^2 = w_A^2 \sigma_A^2 + w_B^2 \sigma_B^2 + 2\rho w_A w_B \sigma_A \sigma_B$, where $w_A = w_B = 1/2$, then $\text{ES}(0.05) = 100000(-\mu_P + \frac{\sigma_P}{0.05} \phi(z_\alpha)) = 24760$, which is smaller than that in (b) because of diversification.