

Q4 (Delta Method)

$$\hat{\theta} = \begin{pmatrix} \hat{\mu} \\ \hat{\sigma}^2 \end{pmatrix} \stackrel{A}{\sim} \mathcal{N}(0, V)$$

$$V = \begin{pmatrix} \sigma^2/n & 0 \\ 0 & 2\sigma^4/n \end{pmatrix}$$

$$= \text{avar}(\hat{\theta})$$

(a) We can consistently estimate  $\text{avar}(\hat{\theta})$  using

$$\hat{V} = \begin{pmatrix} \hat{\sigma}^2/n & 0 \\ 0 & 2(\hat{\sigma}^2)^2/n \end{pmatrix}$$

(b) Let  $q_{r_\alpha} = \mu + \sigma \cdot q_{r_\alpha}^z$ ,  $z \sim N(0,1)$ 

Then

$$\Pr \left( X \leq \mu + \sigma \cdot q_{r_\alpha}^z \right)$$

$$= \Pr\left(\frac{X - \mu}{\sigma} \leq q_{\alpha}^z\right) = \alpha$$

$$\Rightarrow \Pr\left(X \leq \underbrace{\mu + \sigma \cdot q_{\alpha}^z}_{q_{\alpha}^x}\right) = \alpha$$

$$(c) \quad q_{\alpha}^x = g(\theta) = \mu + \sigma \cdot q_{\alpha}^z = \mu + (\sigma^2)^{1/2} q_{\alpha}^z$$

The delta method says

$$\sqrt{n}(g(\hat{\theta}) - g(\theta)) \xrightarrow{d} \mathcal{N}\left(0, \underbrace{\frac{dg(\hat{\theta})}{d\theta} \hat{\Sigma} \frac{dg(\hat{\theta})}{d\theta}}_{\hat{\Sigma}_g}\right)$$

Here

$$\begin{aligned} \frac{dg(\theta)}{d\mu} &= 1, & \frac{dg(\theta)}{d\sigma^2} &= \frac{1}{2} (\sigma^2)^{-1/2} \cdot q_{\alpha}^z \\ & & &= \frac{1}{2} \sigma^{-1} \cdot q_{\alpha}^z \end{aligned}$$

Then

$$V_g = \left( 1, \frac{1}{2} \sigma^{-1} q_{r_2}^2 \right) \begin{bmatrix} \frac{\sigma^2}{n} & 0 \\ 0 & \frac{2\sigma^4}{n} \end{bmatrix} \begin{bmatrix} 1 \\ \frac{1}{2} \sigma^{-1} q_{r_2}^2 \end{bmatrix}$$

$$= \frac{\sigma^2}{n} + \frac{1}{2} \frac{\sigma^2}{n} \cdot (q_{r_2}^2)^2$$

$$= \frac{\sigma^2}{n} \left[ 1 + \frac{1}{2} (q_{r_2}^2)^2 \right]$$

Therefore

$$SE(\hat{q}_{r_2}^x) = \frac{\sigma}{\sqrt{n}} \left[ 1 + \frac{1}{2} (q_{r_2}^2)^2 \right]^{1/2}$$

$$(d) \quad \text{Var}_{0.05} = \$W \times [\exp(q_{r_{0.05}}^x) - 1]$$

$$= W \exp(q_{r_{0.05}}^x) - W$$

$$= g(q_{r_{0.05}}^x)$$

So by delta method

$$\sqrt{n} (g(\hat{q}_{0.05}^x) - g(q_{0.05}^x))$$

$$\xrightarrow{d} N\left(0, \frac{dg(\hat{q}_{0.05}^x)^2}{dq_{0.05}^x} \cdot \widehat{\text{avar}}(\hat{q}_{0.05}^x)\right)$$

$$\frac{dg(q_{0.05}^x)}{dq_{0.05}^x} = W e^{r_f(q_{0.05}^x)} = \text{Var}R_{0.05} + W$$

Therefore

$$\widehat{\text{avar}}(\widehat{\text{Var}}R_{0.05}) = W^2 e^{2r_f(\hat{q}_{0.05}^x)} \cdot \widehat{\text{avar}}(\hat{q}_{0.05}^x)$$

and so

$$SE(\widehat{\text{Var}}R_{0.05}) = W e^{r_f(\hat{q}_{0.05}^x)} \cdot SE(\hat{q}_{0.05}^x)$$

$$(e) \quad SR = \frac{\mu - r_f}{\sigma} = (\sigma^2)^{-1/2} (\mu - r_f)$$

$$\frac{\partial g(\sigma)}{\partial \mu} = \sigma^{-1}, \quad \frac{\partial g(\sigma)}{\partial \sigma^2} = -\frac{1}{2} (\sigma^2)^{-3/2} \times (\mu - r_f)$$

$$= -\frac{1}{2} \sigma^{-3} (\mu - r_f)$$

Then  $\text{avar}(g(\hat{\theta}))$  is

$$\begin{aligned} & \begin{bmatrix} \sigma^{-1} & -\frac{1}{2}\sigma^{-3}(\mu-r_f) \end{bmatrix} \begin{bmatrix} \frac{\sigma^2}{n} & 0 \\ 0 & \frac{2\sigma^4}{n} \end{bmatrix} \begin{bmatrix} \sigma^{-1} \\ -\frac{1}{2}\sigma^{-3}(\mu-r_f) \end{bmatrix} \\ &= \begin{bmatrix} \sigma^{-1} & -\frac{1}{2}\sigma^{-3}(\mu-r_f) \end{bmatrix} \begin{bmatrix} \frac{\sigma}{n} \\ -\frac{\sigma}{n}(\mu-r_f) \end{bmatrix} \end{aligned}$$

$$\begin{aligned} &= \frac{1}{n} + \frac{1}{2} \frac{\sigma^{-2}}{n} (\mu-r_f)^2 = \frac{1}{n} + \frac{1}{2n} SR^2 \\ &= \frac{1}{n} \left[ 1 + \frac{1}{2} SR^2 \right] \end{aligned}$$

Then

$$SE[\hat{SR}] = \frac{1}{\sqrt{n}} \left[ 1 + \frac{1}{2} SR^2 \right]^{1/2}$$