

# CS 476 Assignment 4

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Q5

a)

We have:

$$f(\alpha) = E(\max(L - \alpha, 0))$$

Calculate the expected value using the probability density function  $p(l)$ . Then we have:

$$\begin{aligned} f(\alpha) &= \int_{\alpha}^{\infty} p(L) \cdot (L - \alpha) dL + \int_{-\infty}^{\alpha} p(L) \cdot 0 dL \\ &= \int_{\alpha}^{\infty} p(L) \cdot (L - \alpha) dL \end{aligned}$$

Take derivative on  $f(\alpha)$ . Then we have the following as  $\alpha$  and  $p(L) \cdot (L - \alpha)$  continuously differentiable and using the hint where  $a(x) = x$  and  $g(x, y) = \int_{\alpha}^{\infty} p(x) \cdot (x - y) dx$ .

$$\begin{aligned} \frac{d}{d\alpha} f(\alpha) &= \frac{d}{d\alpha} \left( \int_{\alpha}^{\infty} p(L) \cdot (L - \alpha) dL \right) \\ &= -p(\alpha) \cdot (\alpha - \alpha) \cdot \frac{d}{d\alpha} \alpha + \int_{\alpha}^{\infty} \frac{\partial}{\partial \alpha} (p(L) \cdot (L - \alpha)) dL \\ &= 0 + \int_{\alpha}^{\infty} -p(L) dL \\ &= - \int_{\alpha}^{\infty} p(L) dL \\ &= - \int_{\alpha}^{\infty} p(l) dl \end{aligned}$$

Therefore,  $f(\alpha)$  is continuously differentiable, and  $f'(\alpha) = - \int_{\alpha}^{\infty} p(l) dl$ .

b)

From above, we have  $f'(\alpha) = - \int_{\alpha}^{\infty} p(l) dl$

As  $p(l)$  is a probability distribution function, we have  $\int_{-\infty}^{\infty} p(l) dl = 1$ . As  $\alpha$  increases, then  $\int_{\alpha}^{\infty} p(l) dl$  must decrease. As there exist a negative sign on  $f'(\alpha)$  to  $\int_{\alpha}^{\infty} p(l) dl$ ,  $f'(\alpha)$  is increasing function thus  $f''(\alpha)$  is a positive function for all  $\alpha$ .

As  $f''(\alpha)$  is positive for all  $\alpha$ , the function  $f(\alpha)$  is convex.

c)

Let the equation inside the optimization problem be  $g(\alpha)$ . The equation equals the following:

$$g(\alpha) = \alpha + \frac{1}{1-\beta} \mathbb{E}(\max(L - \alpha, 0)) = \alpha + \frac{1}{1-\beta} f(\alpha)$$

Differentiating the equation gives us the following:

$$g'(\alpha) = \frac{d}{d\alpha} \left( \alpha + \frac{1}{1-\beta} f(\alpha) \right) = 1 - \frac{1}{1-\beta} \int_{\alpha}^{\infty} p(l) dl$$

From b), we already know that  $-\int_{\alpha}^{\infty} p(l) dl$  is an increasing function and 1 is constant, therefore  $g'(\alpha)$  is also increasing. Therefore, the optimization problem is also convex.

To find the minimum value of the optimization problem, it suffices to find the solution of  $g'(\alpha) = 0$ .

$$\begin{aligned} g'(\alpha) &= 0 \\ 1 - \frac{1}{1-\beta} \int_{\alpha}^{\infty} p(l) dl &= 0 \\ \int_{\alpha}^{\infty} p(l) dl &= 1 - \beta \end{aligned}$$

As we are working with the loss  $L$  with a probability distribution with density function  $p(L)$ , by the definition of VaR,  $\alpha^*$ , the solution of the optimization problem, is the VaR with confidence  $\beta$ .