

Assignment 6

Q1

$$u(x) = \frac{x - \alpha x^2}{2}, \quad x \sim N(\mu, \sigma^2)$$

$$\begin{aligned} \Rightarrow \mathbb{E}[u(x)] &= \mathbb{E}\left[\frac{x - \alpha x^2}{2}\right] \\ &= \mathbb{E}\left[\frac{x}{2}\right] - \frac{\alpha}{2} \mathbb{E}[x^2] \\ &= \boxed{\mu - \frac{\alpha}{2} (\sigma^2 + \mu^2)} \end{aligned}$$

\Rightarrow From lecture (via Taylor expansion and $\mathbb{E}[u(x)] = u(x_{CE})$),

$$\begin{aligned} u'(\bar{x}) \cdot (x_{CE} - \bar{x}) &\approx \frac{1}{2} u''(\bar{x}) \cdot \sigma_x^2 \\ \Rightarrow x_{CE} &\approx \bar{x} + \frac{\frac{1}{2} u''(\bar{x}) \sigma_x^2}{u'(\bar{x})} \\ &\approx \mu + \frac{1}{2} \left(\frac{-\alpha \cdot \sigma^2}{1 - \alpha \mu} \right) \\ &\approx \boxed{\mu - \frac{\alpha \sigma^2}{2(1 - \alpha)}} \end{aligned}$$

$u'(x) = 1 - \alpha x$
 $u''(x) = -\alpha$

$$\Rightarrow \pi_A = \bar{\pi} - x_{CE} = \mu - \left(\mu - \frac{\alpha \sigma^2}{2(1 - \alpha)} \right) = \boxed{\frac{\alpha \sigma^2}{2(1 - \alpha)}}$$

Portfolio

$$\Rightarrow W \sim N(1 + r + Z(\mu - r), Z^2 \sigma^2)$$

We need to maximize x_{CE} (same as maximizing utility)

$$x_{CE} \approx 1 + r + Z(\mu - r) - \frac{\alpha Z^2 \sigma^2}{2(1 - \alpha)}$$

$$\frac{\partial x_{CE}}{\partial Z} = \mu - r - \frac{\alpha \sigma^2}{(1 - \alpha)} Z = 0$$

$$\underline{Z^*} = \boxed{\frac{(\mu - r)(1 - \alpha)}{\alpha \sigma^2}}$$

Q2

$\Rightarrow u(x) = \log(x)$ is a special case of $u(x) = \frac{x^{1-\gamma}-1}{1-\gamma}$ when $\gamma = 1$

Therefore,

$$X_{CE} = e^{\mu + \frac{\sigma^2}{2}(1-\gamma)}$$
$$= e^{\mu} \text{ when } \gamma = 1$$

\Rightarrow we have from lecture,

$$\log W \sim N\left(r + \pi(\mu - r) - \frac{\pi^2 \sigma^2}{2}, \pi^2 \sigma^2\right)$$

\Rightarrow we need to maximize X_{CE}

\Rightarrow maximizing mean \nearrow

$$\Rightarrow \text{maximize } r + \pi(\mu - r) - \frac{\pi^2 \sigma^2}{2}$$

$$\text{find critical pt} \Rightarrow \mu - r - \sigma^2 \pi = 0$$

$$\underline{\pi^*} = \boxed{\frac{\mu - r}{\sigma^2}}$$

Q3

Two outcomes for wealth W

$$\Rightarrow f \cdot W_0(1+\alpha) + W_0(1-f) = W_0(1+f\alpha)$$

$$f \cdot W_0(1-B) + W_0(1-f) = W_0(1+fB)$$

Two outcomes for $\log(\text{utility})$ of W

$$\Rightarrow \log(W_0(1+f\alpha))$$

$$\log(W_0(1-fB))$$

$$\Rightarrow E[\log(W)] = p \cdot \log(W_0(1+f\alpha)) + q \cdot \log(W_0(1-fB))$$

$$\Rightarrow \frac{\partial E[\log(W)]}{\partial f} = \frac{p W_0 \alpha}{W_0(1+f\alpha)} + \frac{-q W_0 B}{W_0(1-fB)} = 0$$

$$\Rightarrow \frac{p}{1+f\alpha} = \frac{q}{1-fB}$$

$$\therefore \underline{f^*} = \boxed{\frac{\alpha p - B q}{2\alpha B}}$$

$\rightarrow f^*$ makes intuitive sense. For some α, B , you will bet more ($\uparrow f^*$) if p or probability of positive return (winning) is higher.

The denominator has αB which means it also considers the relationship (relative payout) between positive return and negative return.