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$$\ell_t = \ell_{t-1} + \varepsilon_t$$

$$\chi_t = \phi \chi_{t-1} + \varepsilon_t. \quad \text{Var}(\varepsilon_t) = \sigma^2 = 1$$

$$\chi_t = \underline{\phi \chi_{t-1}} + \varepsilon_t = \ell_t - \ell_{t-1}$$

$$\chi_t = \varepsilon_t = \ell_t - \underline{\ell_{t-1}}$$

$$\varepsilon_{t-1} = \underline{\ell_{t-1}} - \underline{\ell_{t-2}}$$

$$\varepsilon_{t-2} = \underline{\ell_{t-2}} - \underline{\ell_{t-3}}$$

$$\varepsilon_{t-3} = \underline{\ell_{t-3}} - \underline{\ell_{t-4}}$$

$$y_t = \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \varepsilon_{t-3} = \ell_t - \ell_{t-4}$$

$$E[y_t] = E[\varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \varepsilon_{t-3}] = 0$$

$$\text{Cov}(y_t, y_{t-j}), \quad j=0,$$

$$\begin{aligned} \text{Cov}(y_t, y_t) &= \text{Var}(y_t) = \varepsilon_t^2 + \varepsilon_{t-1}^2 + \varepsilon_{t-2}^2 + \varepsilon_{t-3}^2 \\ &= 4. \end{aligned}$$

$$j=1$$

$$\begin{aligned} \text{Cov}(y_t, y_{t-1}) &= E[y_t - \bar{y}][y_{t-1} - \bar{y}] = E[y_t, y_{t-1}] - E[y_t]E[y_{t-1}] \\ &= E[y_t, y_{t-1}] = E[(\varepsilon_t + \dots + \varepsilon_{t-3})(\varepsilon_{t-1} + \dots + \varepsilon_{t-4})] \\ &= E[\varepsilon_{t-1}^2 + \varepsilon_{t-2}^2 + \varepsilon_{t-3}^2] \\ &= 3 \end{aligned}$$

$$j=2 \quad \text{Cov}(y_t, y_{t-2}) = E[\varepsilon_{t-2}^2 + \varepsilon_{t-3}^2] = 2$$

$$j=3 \quad \text{Cov}(y_t, y_{t-3}) = E[\varepsilon_{t-3}^2] = 1$$

$$j=4. \quad \text{Cov}(y_t, y_{t-4}) = 0 \quad j=5, \quad \text{Cov}(y_t, y_{t-5}) = 0.$$

2. Auto Covariance Drops to 0 after 3 lag
Not Correlated With y_{t-i} term.

Hence it is a MAC(3) Model with $AR=0$.

Problem 2

$$1. \text{Var}(R_{t+1}^e) = \beta^2 \text{Var}(x_t) + \text{Var}(\epsilon_{t+1}) \\ = 1 \cdot 0.05^2 + 0.15^2 = 0.025$$

$$\sigma(R_{t+1}^e) = \sqrt{0.025} = 0.158$$

$$2. R^2 = \rho^2 = \left[\frac{\text{Cov}(R_{t+1}^e, x_t)}{\sqrt{\text{Var}(R_{t+1}^e) \text{Var}(x_t)}} \right]^2, \quad \begin{aligned} & \text{Cov}(R_{t+1}^e, x_t) \\ &= E[R_{t+1}^e x_t] - E[R_{t+1}^e] E[x_t] \\ &= E[(a + \beta x_t + \epsilon_{t+1}) x_t] - E[a + \beta x_t + \epsilon_{t+1}] \cdot E[x_t] \\ &= E[\beta x_t^2 + x_t \epsilon_{t+1}] - E[x_t] \cdot 0.05 \\ &= E[x_t^2] + E[x_t \epsilon_{t+1}] - 0.05^2 \\ &= 0.0025 + 0 - 0.05^2 = 0.0025 \end{aligned}$$

$$\text{Var}[x_t] = E[x_t^2] - E^2[x_t]$$

$$E[x_t^2] = 0.05^2 + 0.05^2 = 0.005$$

$$3. \text{ Sharpe Ratio} = \frac{\text{excess return}}{\sigma_{\text{mkt}}} \\ = \frac{0.05}{\sqrt{0.025}} \\ = 0.316 = 31.6\%$$

$$4. \gamma = 40\% \quad \sigma_t(\varepsilon_{t+1}) = 0.15 \\ a=0, \beta=1.$$

$$a_t = \frac{E[a + \beta x_t + \varepsilon_{t+1}]}{\gamma \sigma_t^2[a + \beta x_t + \varepsilon_{t+1}]} \\ = \frac{\beta E[x_t]}{\gamma(\beta^2 \sigma_t^2(x_t) + \sigma_t^2(\varepsilon_{t+1}))} \\ \sigma_t^2[a + \beta x_t + \varepsilon_{t+1}] \\ = \beta^2 \sigma_t^2[x_t] + \sigma_t^2[\varepsilon_{t+1}]$$

if $x_t = 0\%$, $E[x_t] = 0$
 $a_t = 0. \Rightarrow \text{Sharpe Ratio} = \frac{E[R_{t+1}^e]}{\text{GCR}_{t+1}^e} = 0.$

if $x_t = 10\%$

$$a_t = \frac{0.1}{40\% \times 0.15^2} = 1 \quad 100\% \text{ in risky asset.}$$

Sharpe ratio = $\frac{E[R_{t+1}^e]}{\text{GCR}_{t+1}^e} = \frac{0.1}{0.15} = 0.667.$

$$\begin{aligned}
 \text{(a)} \quad E[a + R_{t+1}] &= E\left[\frac{E_t[R_{t+1}^e]}{\gamma \sigma_f^2[R_{t+1}]} R_{t+1}\right] = \frac{E_t^2[R_{t+1}]}{\frac{40}{9} E[\sigma_f^2[R_{t+1}]]} = \frac{0.05^2}{\frac{40}{9} \cdot 0.025} \\
 \sigma_f^2[R_{t+1}] &= \text{Var}(X_t) + \text{Var}(\varepsilon_{t+1}) \\
 &= 0.0025 + 0.15 \\
 &= 0.025
 \end{aligned}$$

$$\begin{aligned}
 \text{(b)} \quad \text{Var}(x_t R_{t+1}) &= E[x_t^2 (X_t^2 + \sigma_f^2(\varepsilon_{t+1}))] - E[x_t X_t]^2 \\
 &= (E[X_t^2 + \sigma_f^2(\varepsilon_{t+1})]) - E[X_t]^2 \times x_t^2 \\
 &= (0.0025 + 0.05^2 + 0.15^2 - 0.05^2) \times 0.45^2 \\
 &= 0.0050625 \\
 \text{Standard deviation} &= \sqrt{0.0050625} \\
 &= 0.07115125.
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(X_t) &= E[X_t^2] - E[X_t]^2 \\
 0.0025 &= E[X_t^2] - 0.05^2 \\
 E[X_t^2] &= 0.0025 + 0.05^2
 \end{aligned}$$

$$\begin{aligned}
 \text{(c)} \quad \text{Sharpe Ratio} &= \frac{0.0225}{0.07115125} = 0.316 \\
 &= 31.6\%
 \end{aligned}$$

$$\begin{aligned}
 \text{(d) (i)} \quad X_t &= -0.05 \text{ or } 0.15 & E[X_t] &= 0.05 \\
 \text{Var}(R_{t+1}) &= \text{Var}(X_t) + \text{Var}(\varepsilon_{t+1}) & \text{Var}(X_t) &= 0.01 \\
 &= 0.01 + 0.15^2 = 0.0325 & \text{Var}(X_t) &= E[X_t^2] - E[X_t]^2 \\
 E[R_{t+1}] &= E[X_t] = 0.05 & E[X_t^2] &= 0.01 + 0.05^2 \\
 & & &= 0.0125
 \end{aligned}$$

$$\begin{aligned}
 R^2 &= P^2 = \left[\frac{\text{Cov}(R_{t+1}^e, X_t)}{\sqrt{\text{Var}(R_{t+1}^e) \text{Var}(X_t)}} \right]^2, \\
 &= \left[\frac{0.01}{\sqrt{0.0325 \cdot 0.001}} \right]^2 \\
 &= 0.30769 \\
 &\quad \text{Cov}(R_{t+1}^e, X_t) \\
 &= E[R_{t+1}^e \cdot X_t] - E[R_{t+1}^e]E[X_t] \\
 &= E[(\alpha + \beta X_t + \varepsilon_{t+1})X_t] - E[\alpha + \beta X_t + \varepsilon_{t+1}] \cdot E[X_t] \\
 &= E[\beta X_t^2 + X_t \varepsilon_{t+1}] - E[X_t] \cdot 0.05 \\
 &= E[X_t^2] - 0.05^2 \\
 &= 0.0125 - 0.05^2 \\
 &\approx 0.01
 \end{aligned}$$

(ii) $X_t = -5\% \text{ or } 15\%$.

$$\begin{aligned}
 a_t &= \frac{E_t[R_{t+1}]}{\gamma \sigma^2[R_{t+1}]} = \frac{0.05}{\frac{40}{9} \cdot 0.0325} \\
 &= 0.3461538
 \end{aligned}$$

$$\begin{aligned}
 E[a_t R_{t+1}] &= 0.3461538 E[R_{t+1}] \\
 &\approx 0.3461538 \times 0.05 \\
 &\approx 0.01730769
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(\alpha_t R_{t+1}) &= E[\alpha_t^2 (X_t^2 + \sigma_t^2(\varepsilon_{t+1}))] - E[\alpha_t X_t]^2 \\
 &= \alpha_t^2 (E[X_t^2] + E[\sigma_t^2(\varepsilon_{t+1})] - E[X_t]^2) \\
 &= 0.346^2 \times (0.0125 + 0.15^2 - 0.05^2) \\
 &\approx 0.0038907
 \end{aligned}$$

$$\text{Sharpe Ratio} = \frac{0.01730769}{\sqrt{0.0038907}} = 0.277$$