

Homework 3

Problem 1: Year-on-year quarterly data and ARMA dynamics

A substantial amount of quantity data, such as earnings, exhibit seasonalities. These can be hard to model. It is therefore common to use so-called Year-on-Year data (e.g., Q1 earnings vs Q1 earnings a year ago, Q2 earnings vs Q2 earnings a year ago, etc). In this problem we will see that such a practice can induce MA-terms due to the overlap in the quarterly year-on-year observations.

Assume the true quarterly log market earnings follow:

$$e_t = e_{t-1} + x_t$$

$$x_t = \phi x_{t-1} + \epsilon_t$$

where $\text{var}(\epsilon_t) = \sigma_\epsilon^2 = 1$ and ϵ is i.i.d. over time t .

The earnings data you are given is year-on-year earnings growth, which in logs is:

$$y_t \equiv e_t - e_{t-4}$$

1. Assume $\phi = 0$. Derive autocovariances of order 0 through 5 for y_t . I.e., $\text{cov}(y_t, y_{t-1})$ for $j = 0, \dots, 5$.

Based on $x_t = \phi x_{t-1} + \epsilon_t$ where $\phi = 0$, we can get $e_t = e_{t-1} + \epsilon_t$ and :

$$\begin{aligned} y_t &= (e_{t-1} + \epsilon_t) - e_{t-4} \\ &= (e_{t-2} + \epsilon_{t-1}) + \epsilon_t - e_{t-4} \\ &= (e_{t-3} + \epsilon_{t-2}) + \epsilon_{t-1} + \epsilon_t - e_{t-4} \\ &= (e_{t-4} + \epsilon_{t-3}) + \epsilon_{t-2} + \epsilon_{t-1} + \epsilon_t - e_{t-4} \\ &= \epsilon_{t-3} + \epsilon_{t-2} + \epsilon_{t-1} + \epsilon_t \\ E(y_t) &= E(\epsilon_{t-3} + \epsilon_{t-2} + \epsilon_{t-1} + \epsilon_t) = 0 \end{aligned}$$

$$\begin{aligned} j = 0: \text{cov}(y_t, y_t) &= E(y_t^2) - E(y_t)E(y_t) \\ &= E(\epsilon_{t-3}^2 + \epsilon_{t-2}^2 + \epsilon_{t-1}^2 + \epsilon_t^2) - 0 \times 0 \\ &= 1 + 1 + 1 + 1 = 4 \end{aligned}$$

$$\begin{aligned} j = 1: \text{cov}(y_t, y_{t-1}) &= E(y_t y_{t-1}) - E(y_t)E(y_{t-1}) \\ &= E(\epsilon_{t-3}^2 + \epsilon_{t-2}^2 + \epsilon_{t-1}^2) - 0 \times 0 \\ &= 1 + 1 + 1 = 3 \end{aligned}$$

$$\begin{aligned} j = 2: \text{cov}(y_t, y_{t-2}) &= E(y_t y_{t-2}) - E(y_t)E(y_{t-2}) \\ &= E(\epsilon_{t-3}^2 + \epsilon_{t-2}^2) - 0 \times 0 \\ &= 1 + 1 = 2 \end{aligned}$$

$$\begin{aligned} j = 3: \text{cov}(y_t, y_{t-3}) &= E(y_t y_{t-3}) - E(y_t)E(y_{t-3}) \\ &= E(\epsilon_{t-3}^2) - 0 \times 0 \\ &= 1 \end{aligned}$$

$$\begin{aligned} j = 4: \text{cov}(y_t, y_{t-4}) &= E(y_t y_{t-4}) - E(y_t)E(y_{t-4}) \\ &= 0 - 0 \times 0 \\ &= 0 \end{aligned}$$

$$\begin{aligned} j = 4: \text{cov}(y_t, y_{t-4}) &= E(y_t y_{t-4}) - E(y_t)E(y_{t-4}) \\ &= 0 - 0 \times 0 \\ &= 0 \end{aligned}$$

2. Assume $\phi = 0$. Determine the number of AR lags and MA lags you need in the ARMA(p,q) process for y_t . Give the associated AR and MA coefficients.

$$\begin{aligned} y_t &= \epsilon_{t-3} + \epsilon_{t-2} + \epsilon_{t-1} + \epsilon_t \\ y_{t-1} &= \epsilon_{t-4} + \epsilon_{t-3} + \epsilon_{t-2} + \epsilon_{t-1} \end{aligned}$$

The red rectangle shows the overlapping between y_t and y_{t-1} . Thus, this demonstrates that y_t has a MA coefficient of $q = 3$ ($MA(3)$).

Since y_t only contains ϵ_t terms, which are all white noise, we can also say that y_t has a AR coefficient of $p = 0$ ($AR(0)$).

Therefore, y_t has an $ARMA(0,3)$ process.

Problem 2: Market-timing and Sharpe ratios

Much of this class is about prediction. In this problem you will derive how market timing can improve the unconditional Sharpe ratio of a fund. The market timing is based on "forecasting regressions" akin to those we undertake in a VAR. However, we are only forecasting one period ahead here.

Assume you have an estimate of expected annual excess market returns for each time t , called x_t . You estimate the regression

$$R_{t+1}^e = \alpha + \beta x_t + \epsilon_{t+1}$$

and obtain $\hat{\alpha} = 0$, $\hat{\beta} = 1$, and $\sigma(\hat{\epsilon}_{t+1}) = 15\%$. Further, the sample mean and standard deviation of x_t are both 5%.

1. Calculate the standard deviation of excess returns based on the information given.

$$\begin{aligned} Var_t[R_{t+1}^e] &= V_t[\alpha + \beta x_t + \epsilon_{t+1}] = \hat{\beta}^2 V_t[x_t] + \sigma_{\epsilon_{t+1}}^2 = 0.05^2 + 0.15^2 = 0.0025 \\ \sigma_{R_{t+1}^e} &= 0.1581 \end{aligned}$$

2. Calculate the R^2 of the regression based on the information given.

$$\begin{aligned} R^2 = \rho^2 &\Rightarrow Cov(R_{t+1}^e, x_t) = E(R_{t+1}^e x_t) - E(R_{t+1}^e)E(x_t) \\ &= E((\alpha + \beta x_t + \epsilon_{t+1})x_t) - E(\alpha + \beta x_t + \epsilon_{t+1}) \times 0.05 \\ &= E(x_t^2) - E(x_t) \times 0.05 \end{aligned}$$

$$\begin{aligned} Var_t(x_t) &= E(x_t^2) - E(x_t)^2 \\ 0.05^2 &= E(x_t^2) - 0.05^2 \\ E(x_t^2) &= 0.005 \end{aligned}$$

$$Cov(R_{t+1}^e, x_t) = 0.005 - 0.05^2 = 0.0025$$

$$\rho = \frac{Cov(R_{t+1}^e, x_t)}{\sigma_{R_{t+1}^e} \sigma_{x_t}} = \frac{0.0025}{0.1581 \times 0.05} = 0.3163 \Rightarrow \rho^2 = 0.1 = R^2$$

3. Calculate the sample Sharpe ratio of excess market returns based on the information given.

$$SR = \frac{\text{excess return}}{SD_{x_t}} = \frac{E(R_{t+1}^e)}{\sigma_{R_{t+1}^e}} = \frac{0.05}{0.1581} = 0.3163$$

4. Recall from investments that myopic investors chooses a fraction of wealth

$$\alpha_t = \frac{E_t[R_{t+1}^e]}{\gamma \sigma_t^2[R_{t+1}^e]}$$

in the risky asset (the market) at each time t , where we assume risk aversion coefficient, γ , equals 40/9. Further, assume that the residuals ϵ_{t+1} are i.i.d., so $\sigma_t(\epsilon_{t+1}) = 15\%$ for all t . Given this, calculate the weight the investor chooses to hold in the risky asset if $x_t = 0\%$ and if $x_t = 10\%$. What is conditional Sharpe ratio in each of these cases?

$$x_t = 0: \quad E_t(R_{t+1}^e) = E(\alpha + \beta x_t + \epsilon_{t+1}) = x_t = 0 \Rightarrow SR = 0, \alpha = 0$$

$$x_t = 10\%: E_t(R_{t+1}^e) = E(\alpha + \beta x_t + \epsilon_{t+1}) = x_t = 0.1 \Rightarrow SR = \frac{0.1}{0.15} = 0.67$$

$$\alpha = \frac{0.1}{\frac{40}{9} \times 0.15^2} = \frac{0.1}{0.1} = 1$$