

# Empirical Methods

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1/28/2020

## Problem 1

We first note that the  $ARMA(1,1)$  is stationary. The reasoning is as follows. First, note that

$$y_t = 0.95y_{t-1} - 0.9\epsilon_{t-1} + \epsilon_t \quad \text{is equivalent to} \quad (1 - 0.95B)y_t = (1 - 0.9B)\epsilon_t.$$

Therefore, the zero of the characteristic equation of the  $AR$  component is

$$\frac{1}{0.95} = \frac{20}{19} > 1.$$

Since we know the series is stationary, we can find its expectation and variance relatively easily. The expected value of  $y_t$  is

$$E[y_t] = E[0.95y_{t-1} - 0.9\epsilon_{t-1} + \epsilon_t] = 0.95E[y_{t-1}].$$

It follows that  $E[y_t] = 0$ . The variance of  $y_t$  is

$$\begin{aligned} Var(y_t) &= Var(0.95y_{t-1} - 0.9\epsilon_{t-1} + \epsilon_t) \\ &= 0.95^2 Var(y_{t-1}) + 0.9^2 Var(\epsilon_{t-1}) + Var(\epsilon_t) - 2(0.95)(0.9)Cov(y_{t-1}, \epsilon_{t-1}) \\ &= 0.9025 Var(y_{t-1}) + 0.81 Var(\epsilon_{t-1}) + Var(\epsilon_t) - 1.71 Cov(y_{t-1}, \epsilon_{t-1}) \\ &= 0.9025 Var(y_{t-1}) + 0.81 Var(\epsilon_{t-1}) + Var(\epsilon_t) - 1.71 Cov(0.95y_{t-2} - 0.9\epsilon_{t-2} + \epsilon_{t-1}, \epsilon_{t-1}) \\ &= 0.9025 Var(y_{t-1}) + 0.81 Var(\epsilon_{t-1}) + Var(\epsilon_t) - 1.71 Var(\epsilon_{t-1}) \\ &= 0.9025 Var(y_t) + 0.10 Var(\epsilon_t) \\ &= 0.9025 Var(y_t) + 0.00025. \end{aligned}$$

In conclusion, we have

$$Var(y_t) = 0.9025 Var(y_t) + 0.00025 \quad \text{implies} \quad Var(y_t) \approx 0.00256410.$$

## Part 1

To find the first order autocorrelation, we can simply find the covariance between  $y_t$  and  $y_{t+1}$  and then divide by the variance of  $y_t$ , due stationarity. To find the covariance, consider

$$\begin{aligned} y_t y_{t+1} &= y_t(0.95y_t - 0.9\epsilon_t + \epsilon_{t+1}) \\ &= 0.95y_t^2 - 0.9y_t\epsilon_t + y_t\epsilon_{t+1} \\ &= 0.95y_t^2 - 0.9(0.95y_{t-1} - 0.9\epsilon_{t-1} + \epsilon_t)\epsilon_t + y_t\epsilon_{t+1} \\ &= 0.95y_t^2 - 0.855y_{t-1}\epsilon_t + 0.81\epsilon_t\epsilon_{t-1} - 0.9\epsilon_t^2 + y_t\epsilon_{t+1}. \end{aligned}$$

Since the expectation of  $y_t$  is zero, it follows that

$$\begin{aligned} Cov(y_t, y_{t+1}) &= E[0.95y_t^2 - 0.855y_{t-1}\epsilon_t + 0.81\epsilon_t\epsilon_{t-1} - 0.9\epsilon_t^2 + y_t\epsilon_{t+1}] \\ &= 0.95E[y_t^2] - 0.855E[y_{t-1}\epsilon_t] + 0.81E[\epsilon_t\epsilon_{t-1}] - 0.9E[\epsilon_t^2] + E[y_t\epsilon_{t+1}] \\ &= 0.95E[y_t^2] - 0.9E[\epsilon_t^2] \\ &\approx 0.95(0.0025641) - 0.9(0.05^2) \\ &\approx 0.000185897. \end{aligned}$$

We conclude that the first order autocovariance is

$$\rho_1 = \frac{Cov(y_t, y_{t+1})}{Var(y_t)} \approx \frac{0.000185897}{0.0025641} \approx 0.0725001.$$

## Part 2

The calculation for the second order autocorrelation is very similar to the first. Consider

$$y_t y_{t+2} = 0.95 y_t y_{t+1} - 0.9 y_t \epsilon_{t+1} + y_t \epsilon_{t+2}.$$

Once again, we utilize the fact that the expectation of  $y_t$  is zero to conclude that

$$Cov(y_t, y_{t+2}) = 0.95 E[y_t y_{t+1}] - 0.9 E[y_t \epsilon_{t+1}] + E[y_t \epsilon_{t+2}] \approx 0.95(0.000185897) \approx 0.000176602.$$

Dividing by  $Var(y_t)$ , we have that

$$\rho_2 = \frac{Cov(y_t, y_{t+2})}{Var(y_t)} \approx \frac{0.000176602}{0.00256410} \approx 0.0688749.$$

The ratio

$$\frac{\rho_2}{\rho_1} \approx \frac{0.0688749}{0.0725001} \approx 0.95.$$

As can be seen in ACF plots of  $ARMA(1, 1)$  models, the first order  $MA$  term distorts the natural tapering off of the autocorrelations caused by the autoregressive component. However, this distortion only lasts for the first autocorrelation; after that we expect to see the autocorrelations tapering off like an  $AR(1)$  model. The ratio shows exactly that since,  $\phi_1 = 0.95$ .

## Part 3

This is no more than simple computation. We have

$$E_t[y_{t+1}] = 0.95 E_t[y_t] - 0.9 E_t[\epsilon_t] + E_t[\epsilon_{t+1}] = 0.95(0.6) - 0.9(0.1) = 0.48.$$

Using our result for the expectation of  $y_{t+1}$ , we have

$$E_t[y_{t+2}] = 0.95 E_t[y_{t+1}] - 0.9 E_t[\epsilon_{t+1}] + E_t[\epsilon_{t+2}] = 0.95(0.48) = 0.456.$$

## Part 4

We have that

$$\hat{x}_t = E_t[y_{t+1}] = 0.95 E_t[y_t] - 0.9 E_t[\epsilon_t] + E_t[\epsilon_{t+1}] = 0.95 y_t - 0.9 \epsilon_t.$$

It follows that

$$E[\hat{x}_t] = 0.95 E[y_t] - 0.9 E[\epsilon_t] = 0.$$

Calculating the variance is a bit more challenging. We know that

$$\begin{aligned} \hat{x}_t &= 0.95 y_t - 0.9 \epsilon_t \\ &= 0.95(0.95 y_{t-1} - 0.9 \epsilon_{t-1} + \epsilon_t) - 0.9 \epsilon_t \\ &= 0.9025 y_{t-1} - 0.855 \epsilon_{t-1} + 0.95 \epsilon_t - 0.9 \epsilon_t \\ &= 0.9025 y_{t-1} - 0.855 \epsilon_{t-1} + 0.05 \epsilon_t. \end{aligned}$$

Furthermore, it is clear

$$\hat{x}_{t-1} = 0.95 y_{t-1} - 0.9 \epsilon_{t-1} \quad \text{implies} \quad 0.95 \hat{x}_{t-1} = 0.9025 y_{t-1} - 0.855 \epsilon_{t-1}.$$

It follows that

$$\hat{x}_t - 0.95 \hat{x}_{t-1} = 0.05 \epsilon_t.$$

We conclude that  $\hat{x}_t$  is, in fact, a stationary  $AR(1)$  model. We are ready to find its variance

$$Var(\hat{x}_t) = 0.95^2 Var(\hat{x}_{t-1}) + 0.05^2 Var(\epsilon_t) \quad \text{implies} \quad Var(\hat{x}_t) = \frac{0.05^4}{1 - 0.95^2} \approx 0.00800641^2.$$

Hence, its standard deviation is

$$\sigma(\hat{x}_t) \approx 0.00800641.$$

All that is left is to find the autocorrelations. Using techniques identical to those in Problem 1.2, we see that

$$Cov(\hat{x}_t, \hat{x}_{t+1}) = E[\hat{x}_t \hat{x}_{t+1}] = 0.95 Var(\hat{x}_t).$$

Therefore, the first order autocorrelation is  $\rho_1 = 0.95$ .

## Problem 2

Observe that

$$y_t = e_t - e_{t-4} = e_{t-1} + \epsilon_t - (e_{t-5} + \epsilon_{t-4}) = e_{t-1} - e_{t-5} + \epsilon_t - \epsilon_{t-4} = y_{t-1} + \epsilon_t - \epsilon_{t-4}.$$

It follows that

$$(1 - B)(y_t - \mu_y) = (1 - B^4)\epsilon_t \quad \text{implies} \quad y_t = \mu_y + \frac{1 - B^4}{1 - B}\epsilon_t = \mu_y + (1 + B + B^2 + B^3)\epsilon_t.$$

In otherwords,

$$y_t = \mu_y + \epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}.$$

Since the  $\epsilon$ -values are independent and identically distributed, it follows that

$$Var(y_t) = Var(\epsilon_t) + Var(\epsilon_{t-1}) + Var(\epsilon_{t-2}) + Var(\epsilon_{t-3}) = 4,$$

when  $t \geq 4$ . As can be seen from the characteristic equation, this series is *not* stationary; for values of  $t$  less than four the variance changes. In particular,  $Var(y_1) = 1$ ,  $Var(y_2) = 2$ , and  $Var(y_3) = 3$ .

## Part 1

Let us find the autocovariances when  $t \geq 3$ :

$$\begin{aligned} Cov(y_t, y_t) &= Var(y_t) \\ Cov(y_t, y_{t+1}) &= Cov(\epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}, \epsilon_{t+1} + \epsilon_t + \epsilon_{t-1} + \epsilon_{t-2}) \\ &= 3, \\ Cov(y_t, y_{t+2}) &= Cov(\epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}, \epsilon_{t+2} + \epsilon_{t+1} + \epsilon_t + \epsilon_{t-1}) \\ &= 2, \\ Cov(y_t, y_{t+3}) &= Cov(\epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}, \epsilon_{t+3} + \epsilon_{t+2} + \epsilon_{t+1} + \epsilon_t) \\ &= 1, \\ Cov(y_t, y_{t+4}) &= Cov(\epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}, \epsilon_{t+4} + \epsilon_{t+3} + \epsilon_{t+2} + \epsilon_{t+1}) \\ &= 0, \end{aligned}$$

and

$$Cov(y_t, y_{t+5}) = Cov(\epsilon_t + \epsilon_{t-1} + \epsilon_{t-2} + \epsilon_{t-3}, \epsilon_{t+5} + \epsilon_{t+4} + \epsilon_{t+3} + \epsilon_{t+2}) = 0.$$

## Part 2

We already completed the reasoning for this step in the preamble for Problem 3. Our conclusion was that

$$y_t = y_{t-1} + \epsilon_t - \epsilon_{t-4}.$$

This is an  $ARMA(1, 4)$  model of the form

$$y_t = \phi_0 + \phi_1 y_{t-1} + \epsilon_t - \theta_1 \epsilon_{t-1} - \dots - \theta_4 \epsilon_{t-4},$$

where

$$\phi_0 = 0, \quad \phi_1 = 1, \quad \theta_1 = 0, \quad \theta_2 = 0, \quad \theta_3 = 0, \quad \text{and} \quad \theta_4 = 1.$$

### Problem 3

#### Part 1

Due to the fact the  $\epsilon_{t+1}$  is independent of  $x_t$ , we have

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

It follows that the standard deviation of excess returns is

$$\sqrt{0.025} \approx 0.158114.$$

#### Part 2

We know that

$$\beta = \rho \frac{\sigma(R_{t+1}^e)}{\sigma(x_t)} \approx 1 \quad \text{implies} \quad R^2 \approx \frac{\sigma^2(x_t)}{\sigma^2(R_{t+1}^e)} = 0.1.$$

#### Part 3

The expected value of excess returns is

$$E[R_{t+1}^2] = E[x_t] = 0.05.$$

Therefore, the Sharpe ratio based on the information given is

$$\frac{0.05}{\sqrt{0.025}} \approx 0.316228.$$

#### Part 4

Let us find  $\alpha_t$ . We know

$$E_t[R_{t+1}^e] = E[x_t] + E_t[\epsilon_{t+1}] = x_t$$

and

$$\text{Var}_t(R_{t+1}^e) = \text{Var}_t(\epsilon_{t+1}) = 0.15^2.$$

It follows that

$$\alpha_t = \frac{x_t}{0.15^2 \gamma} = 10x_t.$$

Hence, if  $x_t$  is equal to 0 or 0.10, then

$$\alpha_t = 0 \quad \text{or} \quad 1,$$

respectively.

Let us find the Sharpe ratio under these two scenarios. If  $x_t = 0$ , then the Sharpe ratio of the portfolio is 0. If  $x_t = 0.10$ , then the Sharpe ratio is

$$\frac{0.10}{0.15} = \frac{2}{3} \approx 0.666667.$$

#### Part 5

- (a) Let us find the unconditional expectation of excess returns under this strategy. Using the law of iterated expectations, we have

$$E[\alpha_t R_{t+1}^e] = E[E_t[\alpha_t R_{t+1}^e]] = E[\alpha_t x_t] = \frac{1}{0.15^2 \gamma} E[x_t^2].$$

Since

$$E[x_t^2] = \frac{1}{2}(0)^2 + \frac{1}{2}(0.10)^2 = 0.005,$$

we conclude that

$$E[\alpha_t R_{t+1}^e] = \frac{0.005}{0.15^2 \gamma} = 0.05.$$

(b) We are given that the variance

$$\text{Var}(\alpha_t R_{t+1}^e) = E \left[ \alpha_t^2 (x_t^2 + \sigma_t^2(\epsilon_{t+1})) \right] - E[\alpha_t x_t]^2.$$

We know that

$$\sigma_t^2(R_{t+1}^e) = \sigma_t^2(\epsilon_{t+1}) = 0.15^2 \quad \text{and} \quad E_t[R_{t+1}^e] = x_t.$$

Then

$$\begin{aligned} E \left[ \alpha_t^2 (x_t^2 + \sigma_t^2(\epsilon_{t+1})) \right] - E[\alpha_t x_t]^2 &= E \left[ \left( \frac{x_t}{0.15^2 \gamma} \right)^2 (x_t^2 + 0.15^2) \right] - E \left[ \left( \frac{x_t}{0.15^2 \gamma} \right) x_t \right]^2 \\ &= \frac{1}{0.15^4 \gamma^2} E [x_t^4 + 0.15^2 x_t^2] - \frac{1}{0.15^4 \gamma^2} E [x_t^2]^2 \\ &= \frac{1}{0.15^4 \gamma^2} (E[x_t^4] + 0.15^2 E[x_t^2] - E[x_t^2]^2) \\ &= \frac{1}{0.15^4 \gamma^2} \left( \frac{0.10^4}{2} + \frac{0.15^2}{2} (0.10)^2 - \frac{1}{4} (0.10)^4 \right) \\ &= 0.0137500. \end{aligned}$$

We conclude that the standard deviation is

$$\sqrt{0.01375} \approx 0.117260.$$

(c) Therefore, the unconditional Sharpe ratio of this strategy is

$$\frac{0.05}{0.117260} \approx 0.426403.$$

(d) Let us find the expected value and variance of  $x_t$  in this situation. The expected value is

$$E[x_t] = -\frac{0.05}{2} + \frac{0.15}{2} = 0.05$$

and the variance is

$$\text{Var}(x_t) = \frac{1}{2}(-0.10)^2 + \frac{1}{2}(0.10)^2 = 0.01.$$

(e) We will proceed to find the  $R^2$  value like we did before. We know

$$\beta = \rho \frac{\sigma(R_{t+1}^e)}{\sigma(x_t)} \approx 1 \quad \text{implies} \quad R^2 \approx \frac{\sigma^2(x_t)}{\sigma^2(R_{t+1}^e)} = \frac{0.01}{0.01 + 0.15^2} \approx 0.308.$$

(ii) To find the Sharpe ratio, we need to find the expectation and variance. The expectation is

$$\begin{aligned} E[\alpha_t R_{t+1}^e] &= E \left[ \frac{E_t[R_{t+1}^e]}{\gamma \sigma_t^2(R_{t+1}^e)} E_t[R_{t+1}^e] \right] \\ &= E \left[ \frac{x_t^2}{0.15^2 \gamma} \right] \\ &= \frac{1}{0.15^2 \gamma} \left[ \frac{1}{2}(-0.05)^2 + \frac{1}{2}(0.15)^2 \right] \\ &= 0.125. \end{aligned}$$

The variance

$$\begin{aligned}
E \left[ \alpha_t^2 \left( x_t^2 + \sigma_t^2(\epsilon_{t+1}) \right) \right] - E[\alpha_t x_t]^2 &= E \left[ \left( \frac{x_t}{0.15^2 \gamma} \right)^2 (x_t^2 + 0.15^2) \right] - E \left[ \left( \frac{x_t}{0.15^2 \gamma} \right) x_t \right]^2 \\
&= \frac{1}{0.15^4 \gamma^2} E [x_t^4 + 0.15^2 x_t^2] - \frac{1}{0.15^4 \gamma^2} E [x_t^2]^2 \\
&= \frac{1}{0.15^4 \gamma^2} (E[x_t^4] + 0.15^2 E[x_t^2] - E[x_t^2]^2) \\
&= \frac{1}{0.15^4 \gamma^2} (0.000256 + 0.15^2(0.0125) - 0.0125^2) \\
&\approx 0.038125.
\end{aligned}$$

We conclude that the standard deviation is

$$\sqrt{0.038125} \approx 0.195256.$$

Hence, the Sharpe ratio is

$$\frac{0.125}{0.195256} \approx 0.640184.$$