應用財務計量經濟學- HW1

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1.4

若 Y_t 為嚴格定態 (strict stationary),說明 Y_t 為具有相同分配之隨機變數。i.e., strict stationary \Longrightarrow identical distributions.

Let $c_1, ..., c_k$ be constants. Then by definition of strict stationarity, we have

$$P(X_{t_1} \leq c_1,...,X_{t_k} \leq c_k) = P(X_{t_1+h} \leq c_1,...,X_{t_k+h} \leq c_k),$$

 $\forall k = 1, 2, ... \text{ and } \forall h = 0, \pm 1, \pm 2, ...$

In particular, we have

$$P(X_{t_1} \leq c_1) = P(X_{t_1+h} \leq c_1)$$

, by choosing $t_k = t_1$. This is true for any t_1 and h.

 \therefore if a sequence is strictly stationary, then all the random variables in the sequence have the same distribution.

2.2

在台灣央行下載外匯存底資料(1987M5~2021M12, 單位: 百萬美元)

(a). 讀入資料, 命名為 FR

```
library(readxl)
fx <- read_excel("Reserves.xlsx")
colnames(fx) <- c("year","FR")

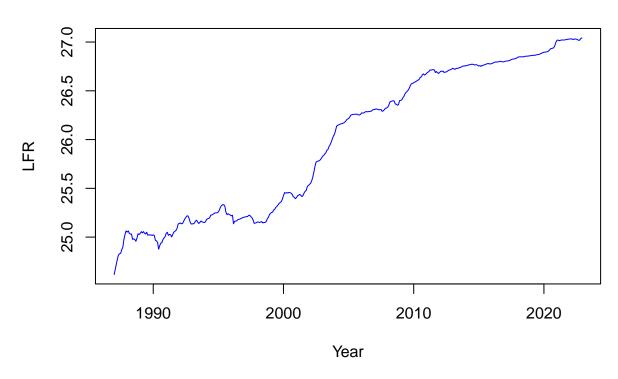
fx$FR <- fx$FR*(10^6) # 轉為單位美元
```

```
fx$year <- NULL # 原先年資料難以操作 (民國年), 所以刪除
# 新建西元年資料
fx$date <- seq(as.Date("1987-01-01"), as.Date("2022-12-01"), by="month")
fx <- fx[order(fx$date),]
```

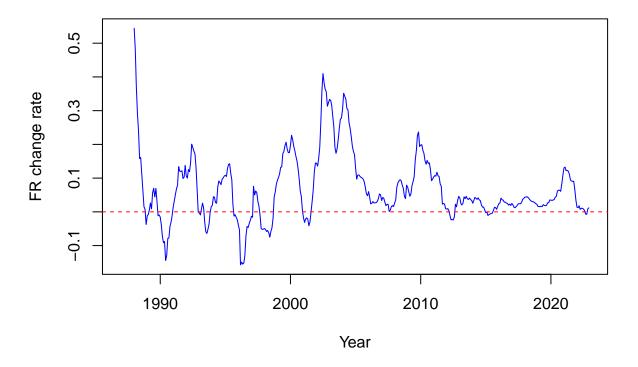
(b). 將 FR 取對數, 命名為 LFR

```
fx$LFR <- log(fx$FR)</pre>
str(fx)
## tibble [432 x 3] (S3: tbl_df/tbl/data.frame)
   $ FR : num [1:432] 4.91e+10 5.18e+10 5.45e+10 5.75e+10 5.99e+10 ...
   $ date: Date[1:432], format: "1987-01-01" "1987-02-01" ...
##
   $ LFR: num [1:432] 24.6 24.7 24.7 24.8 24.8 ...
(c). 計算外匯存底年增率, 命名為 D12FR
library(dplyr, quietly = TRUE)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
fx$prev_year_FR <- lag(fx$FR, 12)</pre>
fx$D12FR <- (fx$FR-fx$prev_year_FR)/fx$prev_year_FR</pre>
(d). 畫出 LFR, D12FR
plot(y=fx$LFR, x=fx$date,type="1", col="blue",
     xlab="Year", ylab="LFR",
    main=expression(log("Taiwan Foreign Reserves")))
```

log(Taiwan Foreign Reserves)



Taiwan Foreign Reserves Change Rate



(e). 建構虛擬變數 Dum

fx\$Dum=0
fx\$Dum <- ifelse(fx\$D12FR>=0,1,0)
mean(fx\$Dum, na.rm=T)

[1] 0.8142857

3.1

 Y_t 是 MA(1) 序列:

$$Y_t = \epsilon_t + \theta \epsilon_{t-1}, \epsilon_t \sim WN(0, \sigma^2)$$

- (a) 計算 $\gamma(j), j \geq 0$
 - if j=0:

$$\begin{split} \gamma(0) &= Cov(Y_t, Y_t) \\ &= Cov(\epsilon_t + \theta \epsilon_{t-1}, \epsilon_t + \theta \epsilon_{t-1}) \\ &= Var(\epsilon_t) + \theta^2 Var(\epsilon_{t-1}) + 2\theta Cov(\epsilon_t, \epsilon_{t-1}) \\ &= \sigma^2 + \theta^2 \sigma^2 + 2\theta [E(\epsilon_t \epsilon_{t-1}) - E(\epsilon_t) E(\epsilon_{t-1})] \\ &= \sigma^2 + \theta^2 \sigma^2 \end{split}$$

• if j=1:

$$\gamma(1) = Cov(Y_t, Y_{t-1}) = Cov(\epsilon_t + \theta \epsilon_{t-1}, \epsilon_{t-1} + \theta \epsilon_{t-2}) = \theta \sigma^2$$

• if j=2:

$$\gamma(1) = Cov(Y_t, Y_{t-2}) = Cov(\epsilon_t + \theta \epsilon_{t-1}, \epsilon_{t-2} + \theta \epsilon_{t-3}) = 0$$

Therefore,

$$\gamma(j) = \begin{cases} \sigma^2 + \theta^2 \sigma^2, & \text{if } j = 0\\ \theta \sigma^2, & \text{if } j = 1\\ 0, & \text{if } j > 1 \end{cases}$$

(b) Y_t 是否定態?

 Y_t 是 weakly stationary, 因為 $E(Y_t)=0, \forall t, Var(Y_t)=\sigma^2+\theta^2\sigma^2<\infty,$ 並且 Autocovariance 只與"j" 有關

事實上, finite-order MA process 都是定態。

3.2

$$\epsilon_t \sim WN(0, \sigma^2)$$

(I)

$$Y_t = 1.2Y_{t-1} - 0.2Y_{t-2} + \epsilon_t = 1.2LY_t - 0.2L^2Y_t + \epsilon_t \implies (1 - 1.2L + 0.2L^2)Y_t = \epsilon_t$$

```
polyroot(z=c(1,-1.2,0.2))
```

```
## [1] 1+0i 5-0i
```

```
abs(polyroot(z=c(1,-1.2,0.2)))
```

[1] 1 5

兩個根分別為1,5,沒有全部大於1,因此非定態。

```
set.seed(123)
e <- rnorm(n = 200, mean = 0, sd = 1)

y <- numeric(length = 200)
y.1 <- 0
y.2 <- 0
phi1 <- 1.2
phi2 <- -0.2

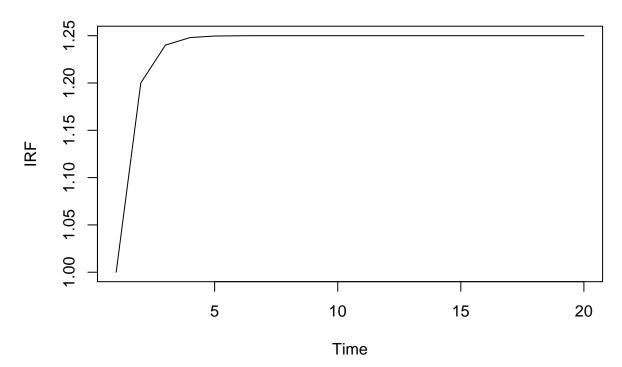
for (i in 1:length(y)) {
    y[i] <- phi1*y.1 + phi2*y.2 + e[i]
    y.2 <- y.1
    y.1 <- y[i]
}

# 計算 IRF

irf <- numeric(length = 20)
irf[1] <- 1
```

```
for (i in 2:length(irf)) {
    if (i == 2) {
        irf[i] <- phi1
    } else {
        irf[i] <- irf[i-1]*phi1 + irf[i-2]*phi2
    }
}

# 繪製 IRF
plot(irf, type = "l", main = "Impulse Response Function",
        xlab = "Time", ylab = "IRF")
```



(II)
$$Y_t = 1.2Y_{t-1} - 0.4Y_{t-2} + \epsilon_t = 1.2LY_t - 0.4L^2Y_t + \epsilon_t \implies (1 - 1.2L + 0.4L^2)Y_t = \epsilon_t$$
 polyroot(z=c(1,-1.2,0.4))

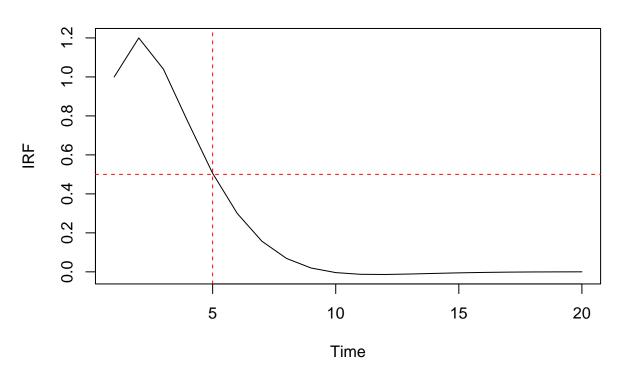
```
## [1] 1.5+0.5i 1.5-0.5i
```

```
abs(polyroot(z=c(1,-1.2,0.4)))
```

```
## [1] 1.581139 1.581139
```

兩個根的絕對值>1,因此定態。

```
set.seed(123)
e <- rnorm(n = 200, mean = 0, sd = 1)
y <- numeric(length = 200)
y.1 <- 0
y.2 < -0
phi1 <- 1.2
phi2 < -0.4
for (i in 1:length(y)) {
  y[i] \leftarrow phi1*y.1 + phi2*y.2 + e[i]
 y.2 < - y.1
 y.1 \leftarrow y[i]
}
# 計算 IRF
irf <- numeric(length = 20)</pre>
irf[1] <- 1
for (i in 2:length(irf)) {
  if (i == 2) {
    irf[i] <- phi1</pre>
  } else {
    irf[i] \leftarrow irf[i-1]*phi1 + irf[i-2]*phi2
  }
}
# 繪製 IRF
plot(irf, type = "1", main = "Impulse Response Function",
     xlab = "Time", ylab = "IRF")
abline(h=0.5, lty=2, col="red")
abline(v=5, lty=2, col="red")
```



```
# 計算半衰期
distance=1
i_min <- 1
for (i in 1:length(irf)){
    if (abs(irf[i]-0.5)<distance){
        distance <- abs(irf[i]-0.5)
        i_min <- i
    }
}
```

[1] 5

```
irf[i_min-1]
```

[1] 0.768

```
irf[i_min]
```

[1] 0.5056

```
irf[i_min+1]
## [1] 0.29952
半衰期約是在 t=5 時。
(III)
        Y_t = 1.2Y_{t-1} + 1.2Y_{t-2} + \epsilon_t = 1.2LY_t + 1.2L^2Y_t + \epsilon_t \implies (1 - 1.2L - 1.2L^2)Y_t = \epsilon_t
polyroot(z=c(1,-1.2,-1.2))
## [1] 0.540833-0i -1.540833+0i
abs(polyroot(z=c(1,-1.2,-1.2)))
## [1] 0.540833 1.540833
其中一根的絕對值 <1, 因此非定態。
set.seed(123)
e < rnorm(n = 200, mean = 0, sd = 1)
y <- numeric(length = 200)
y.1 < -0
y.2 <- 0
phi1 <- 1.2
phi2 <- 1.2
for (i in 1:length(y)) {
  y[i] \leftarrow phi1*y.1 + phi2*y.2 + e[i]
  y.2 < - y.1
```

 $y.1 \leftarrow y[i]$

irf <- numeric(length = 20)</pre>

for (i in 2:length(irf)) {

計算 IRF

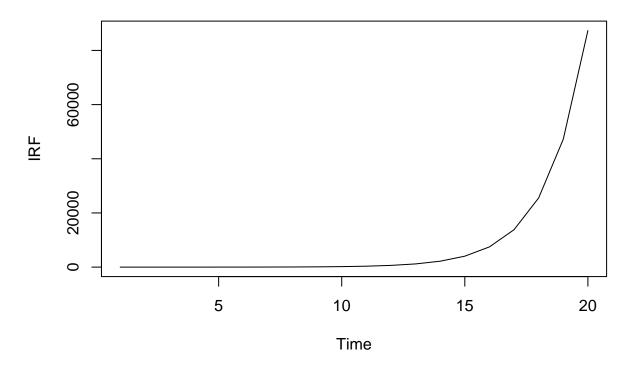
irf[1] <- 1

if (i == 2) {

}

```
irf[i] <- phi1
} else {
    irf[i] <- irf[i-1]*phi1 + irf[i-2]*phi2
}

# 繪製 IRF
plot(irf, type = "l", main = "Impulse Response Function",
    xlab = "Time", ylab = "IRF")</pre>
```



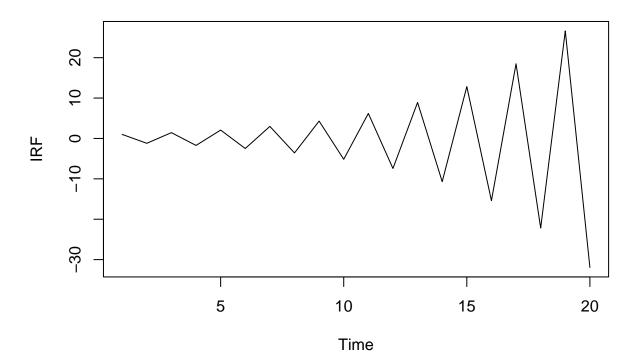
```
(IV) Y_t = -1.2Y_{t-1} + \epsilon_t = -1.2LY_t + \epsilon_t \implies (1+1.2L)Y_t = \epsilon_t
```

```
polyroot(z=c(1,1.2))
## [1] -0.8333333+0i
abs(polyroot(z=c(1,1.2)))
```

```
## [1] 0.8333333
```

根的絕對值 <1, 因此非定態。

```
set.seed(123)
e \leftarrow rnorm(n = 200, mean = 0, sd = 1)
y <- numeric(length = 200)
y.1 <- 0
phi <- -1.2
for (i in 1:length(y)) {
 y[i] \leftarrow phi*y.1 + e[i]
 y.1 < - y[i]
}
# 計算 IRF
irf <- numeric(length = 20)</pre>
irf[1] <- 1
for (i in 2:length(irf)) {
 irf[i] <- irf[i-1]*phi</pre>
}
# 繪製 IRF
plot(irf, type = "1", main = "Impulse Response Function",
     xlab = "Time", ylab = "IRF")
```



(V)

$$\begin{split} Y_t &= 0.7Y_{t-1} + 0.25Y_{t-2} - 0.175Y_{t-3} + \epsilon_t \\ &= 0.7LY_t + 0.25L^2Y_t - 0.175L^3Y_t + \epsilon_t \\ \Longrightarrow & (1 - 0.7L - 0.25L^2 + 0.175L^3)Y_t = \epsilon_t \end{split}$$

polyroot(z=c(1,-0.7,-0.25,0.175))

[1] 1.428571-0i -2.000000+0i 2.000000+0i

```
abs(polyroot(z=c(1,-0.7,-0.25,0.175)))
```

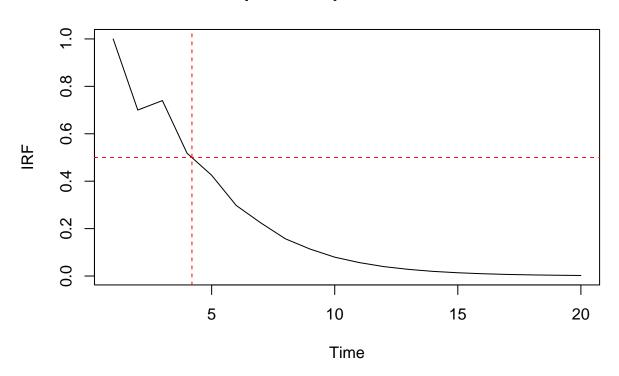
[1] 1.428571 2.000000 2.000000

根的絕對值>1,因此定態。

```
set.seed(123)
e <- rnorm(n = 200, mean = 0, sd = 1)

y <- numeric(length = 200)</pre>
```

```
y.1 < -0
y.2 <- 0
y.3 <- 0
phi1 <- 0.7
phi2 <- 0.25
phi3 <- -0.175
for (i in 1:length(y)) {
  y[i] \leftarrow phi1*y.1 + phi2*y.2+ phi3*y.3 + e[i]
  y.3 < - y.2
 y.2 < - y.1
  y.1 \leftarrow y[i]
}
# 計算 IRF
irf <- numeric(length = 20)</pre>
irf[1] <- 1
irf[2] <- phi1</pre>
for (i in 3:length(irf)) {
  if(i == 3){
    irf[i] \leftarrow irf[i-1]*phi1+phi2
  } else {
    irf[i] \leftarrow irf[i-1]*phi1 + irf[i-2]*phi2+ irf[i-3]*phi3
  }
}
# 繪製 IRF
plot(irf, type = "l", main = "Impulse Response Function",
     xlab = "Time", ylab = "IRF")
abline(h=0.5, lty=2, col="red")
abline(v=4.2, lty=2, col="red")
```



```
# 計算半衰期
distance=1
i_min <- 1
for (i in 1:length(irf)){
    if (abs(irf[i]-0.5)<distance){
        distance <- abs(irf[i]-0.5)
        i_min <- i
    }
}
i_min
```

```
## [1] 4
```

```
irf[i_min-1]
```

```
## [1] 0.74
```

```
irf[i_min]
```

```
## [1] 0.518
```

irf[i_min+1]

[1] 0.4251

半衰期約發生在 t=4 時。

3.3

AR(1):

$$Y_t = \phi_0 + \phi_1 Y_{t-1} + \epsilon_t, \epsilon_t \sim i.i.d.(0, \sigma^2), |\phi_1| < 1, 0 \le t < \infty$$

initial value: Y_0 be a constant

若此 AR(1) 模型是定態,則 $E(Y_t) = \mu, \forall 0 \le t < \infty$

$$\begin{split} E(Y_t) &= E(\phi_0 + \phi_1 Y_{t-1} + \epsilon_t) \\ &= \phi_0 + \phi_1 E(Y_{t-1}) + E(\epsilon_t) = \\ &= \phi_0 + \phi_1 E(Y_{t-1}) \\ &\Longrightarrow \mu = \phi_0 + \phi_1 \mu \\ &\Longrightarrow \mu = \frac{\phi_0}{1 - \phi_1} \end{split}$$

In general, $E(Y_0) = Y_0 \neq \frac{\phi_0}{1-\phi_1}$, 另外, $Var(Y_0) = \gamma(Y_0, Y_0) = 0$,

$$\begin{split} Var(Y_t) &= Var(\phi_0 + \phi_1 Y_{t-1} + \epsilon_t) \\ &= \phi_1^2 Var(Y_{t-1}) + \sigma^2 \\ &\implies \sigma_Y^2 = \phi_1^2 \sigma_Y^2 + \sigma^2 \\ &\implies \sigma_Y^2 = \frac{\sigma^2}{1 - \phi_1^2} \neq 0, \forall t > 0 \end{split}$$

 $Cov(Y_0, Y_{0+h}) = 0, \forall h$

因此, 若 AR(1) 有一固定起始值, 則非定態。

3.4

若 Y_0 是隨機變數,則要滿足

•
$$E(Y_0) = \frac{\phi_0}{1 - \phi_1}$$

$$\begin{array}{l} \bullet \ E(Y_0) = \frac{\phi_0}{1-\phi_1} \\ \bullet \ Var(Y_0) = \frac{\sigma^2}{1-\phi_1^2} \end{array}$$

•
$$\gamma(h) = Cov(Y_0, Y_{0+h}), \forall h$$

才會至少是弱定熊

3.19

(a). AR(1):

$$Y_t = 0.5 - 0.95 Y_{t-1} + \epsilon_t, \epsilon_t \sim i.i.d.N(0,1)$$

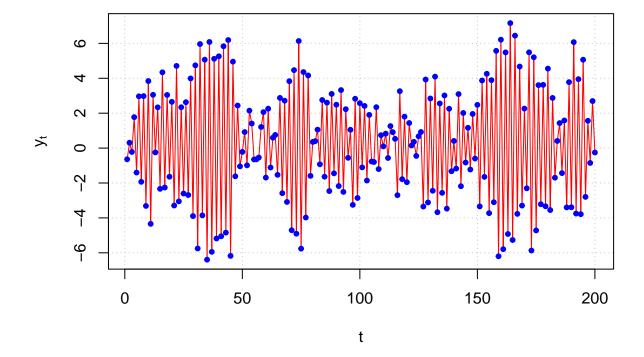
```
set.seed(7181)
e <- rnorm(n = 200, mean = 0, sd = 1)

y <- numeric(length = 200)
y.1 <- 0
for (i in 1:length(y)) {
    y[i] <- 0.5-0.95*y.1 + e[i]
        y.1 <- y[i]
}

plot(x = y, ylab = expression(y[t]), xlab = "t", type =
    "l", col = c("red"), main =
    expression(paste(y[t] == 0.5-0.95*y[t-1] + e[t], " where ",
    e[t], "~ iid. N(0,1)" )), panel.first=grid(col =
    "gray", lty = "dotted"))

points(x = y, pch = 20, col = "blue")</pre>
```

$$y_t = 0.5 - 0.95y_{t-1} + e_t$$
 where $e_t \sim iid. N(0,1)$



(b). MA(1):

$$Y_t = \epsilon_t - 0.95\epsilon_{t-1}, \epsilon_t \sim i.i.d.N(0,1)$$

```
set.seed(8199)
y <- arima.sim(list(ma=c(-0.95)), n=100, rand.gen=rnorm, sd=1)

plot(x = y, ylab = expression(y[t]), xlab = "t", type =
    "l", col = c("red"), main =
    expression(paste(y[t] == e[t] - 0.95*e[t-1], " where ",
    e[t], " ~ iid. N(0,1)")) , panel.first=grid(col =
    "gray", lty = "dotted"))

points(x = y, pch = 20, col = "blue")</pre>
```

 $y_t = e_t - 0.95e_{t-1}$ where $e_t \sim iid. N(0,1)$

