

3.15. For AR(1), $x_t = \phi x_{t-1} + w_t$.

$$x_{t+1} = \phi x_t + w_{t+1}$$

$$x_{t+1}^{(tt)} = \mathbb{E}(x_{t+1} | x_1, \dots, x_t) = \phi x_t$$

$$\begin{aligned} x_{t+2}^{(tt)} &= \mathbb{E}(\phi x_{t+1} + w_{t+2} | x_1, \dots, x_t) \\ &= \mathbb{E}(\phi^2 x_t + \phi w_{t+1} + w_{t+2} | x_1, \dots, x_t) \\ &= \phi^2 x_t \end{aligned}$$

⋮

$$x_{t+m}^{(tt)} = \phi^m x_t.$$

$$\mathbb{E}(x_{t+m} - x_{t+m}^{(tt)})^2 = \mathbb{E}(\phi^m x_t + \phi^{m-1} w_{t+1} + \dots + w_{t+m} - \phi^m x_t)^2$$

$$= \mathbb{E}(\phi^{m-1} w_{t+1} + \phi^{m-2} w_{t+2} + \dots + w_{t+m})^2$$

$$= \sum_{i=0}^{m-1} \mathbb{E}(\phi^{2i} w_{t+m-i}^2) + 2 \sum_{i > k} \phi^{i+k} \mathbb{E}(w_i w_k)$$

Since $\mathbb{E}(w_i w_k) = 0$, $\mathbb{E}(w_i^2) = \sigma_w^2$

$$= \sum_{i=0}^{m-1} \phi^{2i} \cdot \sigma_w^2 = \sigma_w^2 \cdot \frac{1 - \phi^{2m}}{1 - \phi^2}$$

3.21

$$X_t = 0.9 X_{t-1} + W_t + 0.5 W_{t-1} \quad W_t \sim N(0, 1)$$

$$\text{Let } \beta = (\mu, \phi, \theta).$$

① Unconditional MLE:

$$L(\beta, \sigma_w^2) = f(x_1) \cdot \prod_{t=2}^n f(x_t | x_{1:n-1})$$

where $x_t - \mu = \sum_{j=0}^{\infty} \psi_j W_{t-j}$,

$$W_t \stackrel{iid}{\sim} N(0, \sigma_w^2) \quad x_1 \sim N(\mu, \sigma_w^2 \sum_{j=0}^{\infty} \psi_j^2)$$

$$E(x_n | x_{1:n-1}) = x_n^{*}$$

$$\begin{aligned} \text{Var}(x_n | x_{1:n-1}) &= E(x_n - x_n^{*})^2 \\ &= r(n) \prod_{j=1}^{n-1} (1 - \phi_j^2) \end{aligned}$$

$$\begin{aligned} &= \sigma_w^2 \sum_{j=0}^{\infty} \psi_j^2 \left[\prod_{j=1}^{n-1} (1 - \phi_j^2) \right] \\ &= \sigma_w^2 r_n \quad \text{as } r_n \triangleq \sum_{j=0}^{\infty} \psi_j^2 \left[\prod_{j=1}^{n-1} (1 - \phi_j^2) \right] \end{aligned}$$

$$\text{Let } x_0 = \mu, \quad r_1 = \sum_{j=0}^{\infty} \psi_j^2$$

$$\Rightarrow x_t \sim N(\mu, \sigma_w^2 r_t)$$

The likelihood function is given by:

$$L(\beta, \sigma_w^2) = \prod_{t=1}^n \left[2\pi \sigma_w^2 r_t(\beta) \right]^{-\frac{1}{2}} \cdot \exp \left\{ -\frac{(x_t - x_t^{*}(\beta))^2}{2\sigma_w^2 r_t(\beta)} \right\}$$

$$\text{Let } S(\beta) = \sum_{t=1}^n \frac{(x_t - x_t^{*}(\beta))^2}{r_t(\beta)}$$

$$\Rightarrow L(\beta, \sigma_w^2) = (2\pi \sigma_w^2)^{-\frac{n}{2}} [r_1(\beta) \cdots r_n(\beta)]^{-\frac{1}{2}} \cdot \exp \left\{ -\frac{S(\beta)}{2\sigma_w^2} \right\}$$

$$\text{log-likelihood: } \ell(\beta, \sigma_w^2) = \log L(\beta, \sigma_w^2)$$

$$= \log \left[\frac{1}{n} S(\beta) \right] + \frac{1}{n} \sum_{t=1}^n \log r_t(\beta)$$

② Unconditional Least Square:

$$\text{minimize } S(\beta) = \sum_{t=1}^n \frac{(x_t - x_t^{t-1}(\beta))^2}{r_t(\beta)}$$

③ Conditional Least Square.

Condition: x_1, w_1 are non-random

$$\text{Let } w_t(\beta) = x_t - \phi x_{t-1} - \theta w_{t-1}(\beta)$$

Since when $t-1 \geq 1$, x_t, w_t are non-random
 $w_t(\beta)$ can be calculated iteratively.

Want to minimize:

$$S_{cm}(\beta) = \sum_{t=2}^n w_t^2(\beta)$$

STA4003_HW5_codes

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```
library(tswge)
```

```
## Warning: 程辑包'tswge'是用R版本4.2.3 来建造的
```

```
## Registered S3 method overwritten by 'quantmod':  
##   method             from  
##   as.zoo.data.frame zoo
```

```
##  
## 载入程辑包: 'tsvge'
```

```
## The following object is masked from 'package:datasets':  
##  
##      uspop
```

```
set.seed(123)  
  
x <- list()  
  
for(i in 1:10){  
  x[[i]] <- gen.arima.wge(200, phi=0.9, theta=0.5, d=0, s=0, mu=0, vara=1, plot=F, sn=0)  
  print(arima(x[[i]], order = c(1,0,1), method = "CSS-ML"))  
  print(arima(x[[i]], order = c(1,0,1), method = "ML"))  
  print(arima(x[[i]], order = c(1,0,1), method = "CSS"))  
  cat("-----\n")  
}
```

```

##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9144  -0.4367   -0.7202
## s.e.    0.0370   0.0818    0.4019
##
## sigma^2 estimated as 0.8196:  log likelihood = -264.39,  aic = 536.79
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9143  -0.4366   -0.7202
## s.e.    0.0370   0.0818    0.4017
##
## sigma^2 estimated as 0.8196:  log likelihood = -264.39,  aic = 536.79
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9124  -0.4444   -0.9037
## s.e.    0.0365   0.0811    0.4139
##
## sigma^2 estimated as 0.8158:  part log likelihood = -263.42
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8899  -0.4417   -0.2143
## s.e.    0.0424   0.0823    0.3589
##
## sigma^2 estimated as 1.074:  log likelihood = -291.32,  aic = 590.64
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8899  -0.4416   -0.2145
## s.e.    0.0424   0.0823    0.3588
##
## sigma^2 estimated as 1.074:  log likelihood = -291.32,  aic = 590.64
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##

```

```

## Coefficients:
##          ar1          mal  intercept
##          0.8784 -0.4156    -0.0992
## s.e.   0.0449   0.0812    0.3563
##
## sigma^2 estimated as 1.077:  part log likelihood = -291.18
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8503 -0.5200    -0.3946
## s.e.   0.0563   0.0878    0.2099
##
## sigma^2 estimated as 0.8961:  log likelihood = -273.03,  aic = 554.06
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8502 -0.5199    -0.3947
## s.e.   0.0564   0.0878    0.2099
##
## sigma^2 estimated as 0.8961:  log likelihood = -273.03,  aic = 554.06
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8544 -0.5226    -0.3866
## s.e.   0.0568   0.0883    0.2211
##
## sigma^2 estimated as 0.901:  part log likelihood = -273.36
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8614 -0.5485    -0.3148
## s.e.   0.0579   0.0944    0.2511
##
## sigma^2 estimated as 1.243:  log likelihood = -305.77,  aic = 619.54
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8613 -0.5485    -0.3150
## s.e.   0.0579   0.0944    0.2511

```

```

##
## sigma^2 estimated as 1.243:  log likelihood = -305.77,  aic = 619.54
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8651  -0.5531   -0.3527
## s.e.    0.0569   0.0933    0.2630
##
## sigma^2 estimated as 1.248:  part log likelihood = -305.93
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9018  -0.5883    0.4698
## s.e.    0.0444   0.0853    0.2701
##
## sigma^2 estimated as 0.8932:  log likelihood = -272.79,  aic = 553.59
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9018  -0.5882    0.4696
## s.e.    0.0444   0.0854    0.2700
##
## sigma^2 estimated as 0.8932:  log likelihood = -272.79,  aic = 553.59
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8953  -0.5677    0.5234
## s.e.    0.0481   0.0885    0.2801
##
## sigma^2 estimated as 0.9011:  part log likelihood = -273.38
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9323  -0.6371    0.2527
## s.e.    0.0357   0.0815    0.3534
##
## sigma^2 estimated as 0.9589:  log likelihood = -279.97,  aic = 567.93
##
## Call:

```

```

## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9323  -0.6372    0.2531
## s.e.    0.0357   0.0815    0.3535
##
## sigma^2 estimated as 0.9589:  log likelihood = -279.97,  aic = 567.93
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9285  -0.6178    0.1801
## s.e.    0.0389   0.0862    0.3751
##
## sigma^2 estimated as 0.9723:  part log likelihood = -280.98
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9000  -0.5727   -0.6953
## s.e.    0.0437   0.0815    0.2711
##
## sigma^2 estimated as 0.8602:  log likelihood = -269.04,  aic = 546.07
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9001  -0.5728   -0.6951
## s.e.    0.0437   0.0815    0.2712
##
## sigma^2 estimated as 0.8602:  log likelihood = -269.04,  aic = 546.07
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.9009  -0.5732   -0.6326
## s.e.    0.0439   0.0810    0.2844
##
## sigma^2 estimated as 0.8627:  part log likelihood = -269.02
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept

```



```

##      0.8090  -0.3083   -0.1806
## s.e.  0.0585   0.0919    0.2427
##
## sigma^2 estimated as 0.9316:  log likelihood = -276.99,  aic = 561.99
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##      ar1      mal  intercept
##      0.8090  -0.3082   -0.1805
## s.e.  0.0585   0.0919    0.2426
##
## sigma^2 estimated as 0.9316:  log likelihood = -276.99,  aic = 561.99
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##      ar1      mal  intercept
##      0.8121  -0.3111   -0.1475
## s.e.  0.0585   0.0921    0.2513
##
## sigma^2 estimated as 0.9345:  part log likelihood = -277.01
## -----
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##      ar1      mal  intercept
##      0.8695  -0.5313   -0.5668
## s.e.  0.0480   0.0755    0.2429
##
## sigma^2 estimated as 0.9653:  log likelihood = -280.51,  aic = 569.01
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##      ar1      mal  intercept
##      0.8694  -0.5313   -0.5667
## s.e.  0.0480   0.0755    0.2429
##
## sigma^2 estimated as 0.9653:  log likelihood = -280.51,  aic = 569.01
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##      ar1      mal  intercept
##      0.8636  -0.5209   -0.5021
## s.e.  0.0495   0.0755    0.2460
##
## sigma^2 estimated as 0.9669:  part log likelihood = -280.43
## -----

```

```

##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS-ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8820  -0.4946      0.2819
## s.e.    0.0459   0.0828      0.2890
##
## sigma^2 estimated as 0.9647:  log likelihood = -280.51,  aic = 569.02
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "ML")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8820  -0.4946      0.2816
## s.e.    0.0459   0.0828      0.2890
##
## sigma^2 estimated as 0.9647:  log likelihood = -280.51,  aic = 569.02
##
## Call:
## arima(x = x[[i]], order = c(1, 0, 1), method = "CSS")
##
## Coefficients:
##          ar1          mal  intercept
##          0.8842  -0.4977      0.2208
## s.e.    0.0456   0.0821      0.3029
##
## sigma^2 estimated as 0.9674:  part log likelihood = -280.47
## -----

```