

Autoregressive processes II.

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Stationarity of a given AR(p) process

Lecture: AR(1) and AR(2) processes

For the process

$$x_t = \delta + \alpha x_{t-1} + u_t$$

we have shown that stationarity is equivalent to $|\alpha| < 1$.

Writing the process in terms of the lag operator L :

$$(1 - \alpha L)x_t = \delta + u_t$$

Stationarity condition: the root of

$$1 - \alpha L = 0$$

are in absolute value greater than 1.

Lecture: AR(1) and AR(2) processes

The process

$$x_t = \delta + \alpha_1 x_{t-1} + \alpha_2 x_{t-2} + u_t$$

can be written as

$$(1 - \alpha_1 L - \alpha_2 L^2)x_t = \delta + u_t$$

Stationarity condition: the roots of

$$1 - \alpha_1 L - \alpha_2 L^2 = 0$$

are in absolute value greater than 1, i.e., they are **outside of the unit circle**.

General AR(p) process

The process

$$x_t = \delta + \alpha_1 x_{t-1} + \dots + \alpha_p x_{t-p} + u_t$$

can be written as

$$(1 - \alpha_1 L - \dots - \alpha_p L^p) x_t = \delta + u_t$$

Stationarity condition: the roots of

$$1 - \alpha_1 L - \dots - \alpha_p L^p = 0$$

are in absolute value greater than 1, i.e., they are **outside of the unit circle**.

Stationarity in R

We will use the library `fArma` and the function `armaRoots`

```
library(fArma)
```

Roots of

$$1 - \alpha_1 L - \alpha_2 L^2 - \dots - \alpha_p L^p = 0$$

can be found by:

```
armaRoots(c(alfa_1, alfa_2, ..., alfa_p))
```

Stationarity in R

For example:

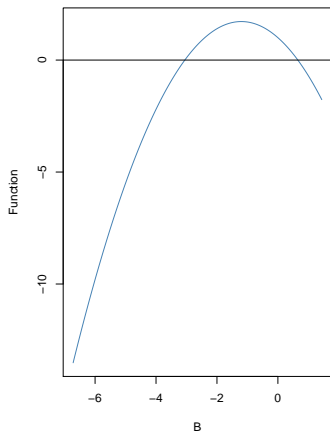
```
# in we want both graphs in the same figure  
# (otherwise they will be in separate figures):  
par(mfrow = c(1, 2))  
  
# roots of 1 - 1.2 L - 0.5 L^2 = 0  
armaRoots(c(1.2,0.5))  
  
# restore original settings  
par(mfrow = c(1, 1))
```

The output is graphical as well as given in a table (real part, imaginary part, absolute value)

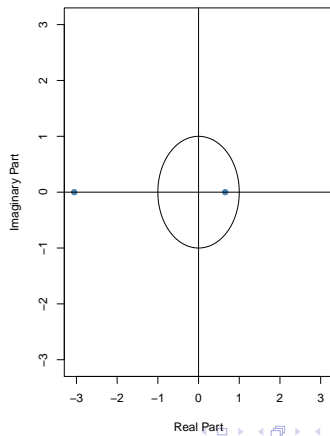
Stationarity in R

Graphically:

Polynomial Function vs. B



Roots and Unit Circle



Stationarity in R

One of the roots is not outside of the unit circle, therefore the process

$$(1 - 1.2L - 0.5L^2)x_t = \delta + u_t$$

is not stationary.

Stationarity in R

Exercises:

- ▶ Check the stationarity of the AR(2) model for the spread from last week lesson
- ▶ Find another example of a stationary AR(2) process
- ▶ Find an example of a nonstationary AR(2) process.
- ▶ Part of *fundamentals* on exam: find out whether the given process is stationary

Autocorrelation function of an AR process

Solving the difference equation

We will solve the difference equations for ACF of the two $AR(2)$ processes from the lecture.

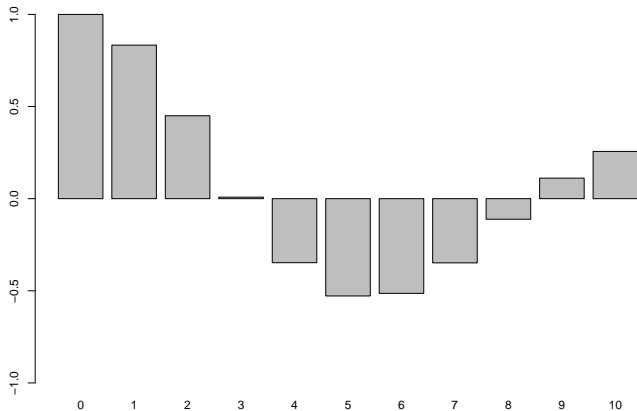
Function ARMAacf

For the process $x_t = 1.5x_{t-1} - 0.8x_{t-2} + u_t$:

```
r <- ARMAacf(ar=c(1.5, -0.8), lag.max=10)
```

Plot the ACF (for example using `barplot`)

Result



Exercise: Modelling long-term unemployment

Data

```
library(WDI)  
WDIsearch('long.*term.*unemployment')
```

```
##      indicator  
## [1,] "SL.UEM.LTRM.FE.ZS"  
## [2,] "SL.UEM.LTRM.MA.ZS"  
## [3,] "SL.UEM.LTRM.ZS"  
##      name  
## [1,] "Long-term unemployment, female (% of female unemp  
## [2,] "Long-term unemployment, male (% of male unemployme  
## [3,] "Long-term unemployment (% of total unemployment)"
```


Data

```
data <- WDI(indicator='SL.UEM.LTRM.ZS',  
            country=c('CA'),  
            start=1982)  
  
data <- data[order(data$year),] # 20 years  
y <- data$SL.UEM.LTRM.ZS  
y <- ts(y, start=1982, frequency=1)
```

Exercise

Find a suitable AR model for y . Check its stationary and residuals.
Does it have a periodical character (as suggested by the plot)?

