

EDHEC PhD Finance 2022 - Econometrics Homework

Giovanni Maffei, Timo Predoehl

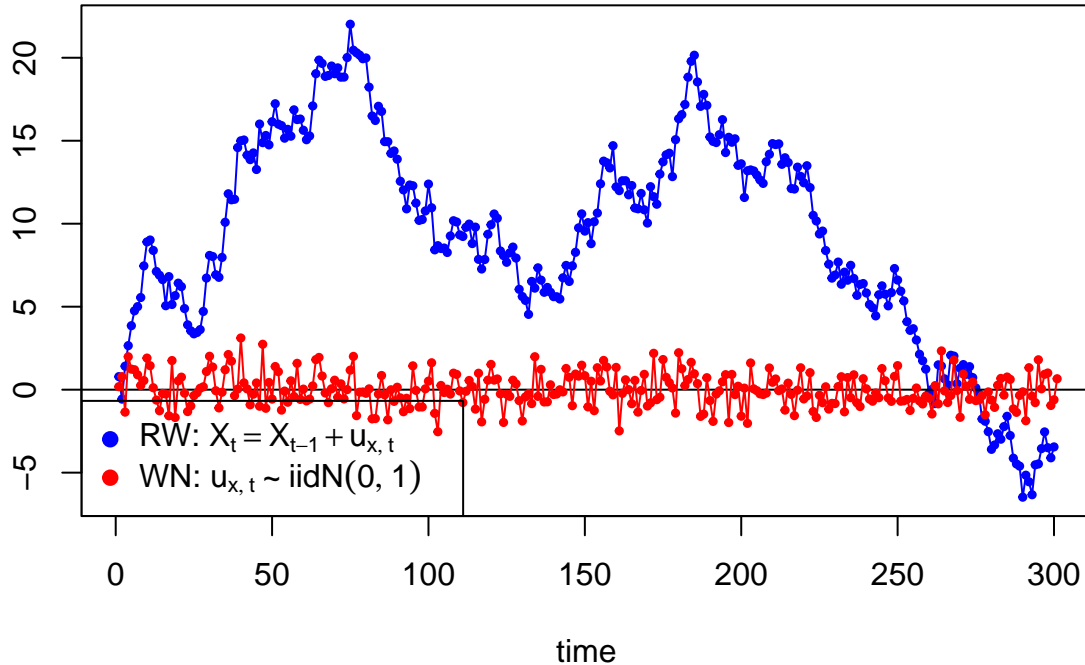
Question 1: Spurious Regressions (2 points)

- **compulsory** (a) [1 Point]
 - i Replicate the analysis leading to Figure 14.1 in Davidson MacKinnon (2005, book) by running 100,000 MC, or simply 10,000 MC if your computer is slow (instead of 1 million), and using $T = 6, 12, 60, 120, 240, 360, 480$. (i) Compute also for each sample size T the distribution of the R^2 s of the MC simulations with either 7 separate histograms, or one unique figure where you report on the y -axis the 5%, 10%, 25%, 50%, 75%, 90% and 95% quantiles of the distributions of the simulated R^2 , and on the x -axis you have $T = 6, 12, 60, 120, 240, 360, 480$.
 - ii Similarly (either with histograms, or with one plot of the quantiles) report the distributions of the estimates t-statistics for the test of the null $H_0 : \beta_2 = 0$ and
 - iii their empirical rejection frequencies (that is the empirical size of the tests), which is exactly the figure 14.1 in Davidson MacKinnon (2005, book).
- (b) [1 Point] Based on the results obtained by answering to point (a) summarize the problems of spurious regressions in econometrics.

(a)(i) Replication Davidson McKinnon, 2005, fig 14.1

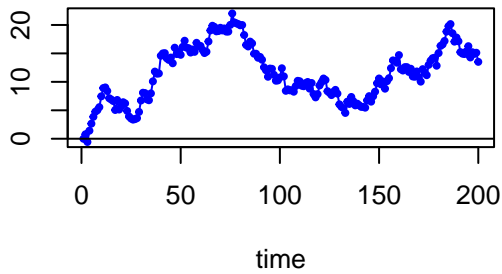
Simulate a random walk processes $y_t = y_{t-1} + e_t, y_0 = 0, e_t \sim IID(0, \sigma^2)$ analog to (14.03)

White Noise and Random Walk

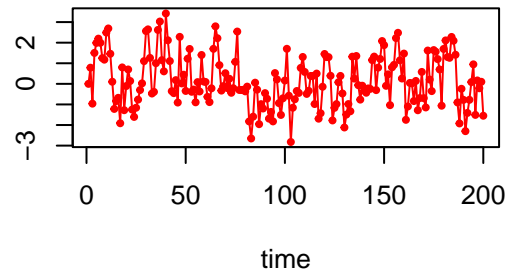


Simulate two AR(1) processes according to $x_t = \phi x_{t-1} + u_t$, with $\phi = 0.5$ and with $\phi = 1$. The ACFs of two processes indicate that in latter significant autocorrelation persists at least up to 20 lags, while the former autocorrelation becomes insignificant after 3 lags. Hence, in the case of AR(1), autocorrelation is a function of the parameter ϕ .

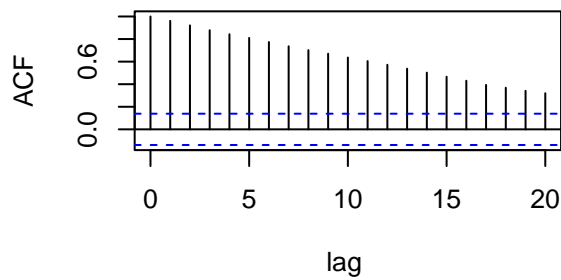
AR(1): $X_t = X_{t-1} + u_{x,t}$, $u_{x,t} \sim \text{iidN}(0, 1)$



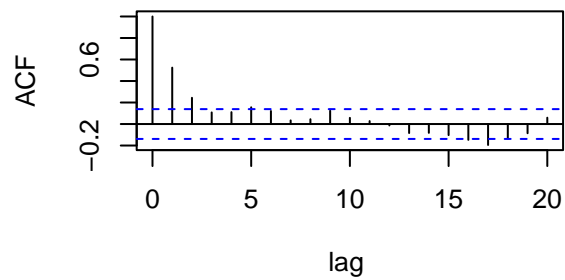
AR(1): $X_t = 0.5X_{t-1} + u_{x,t}$, $u_{x,t} \sim \text{iidN}(0, 1)$



ACF, AR(1) NON-stationary



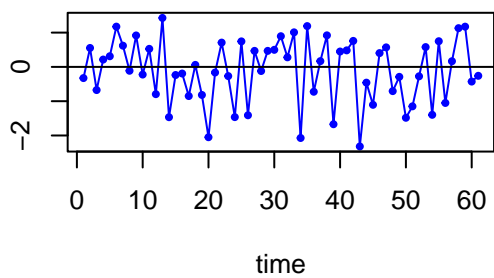
ACF, AR(1) stationary



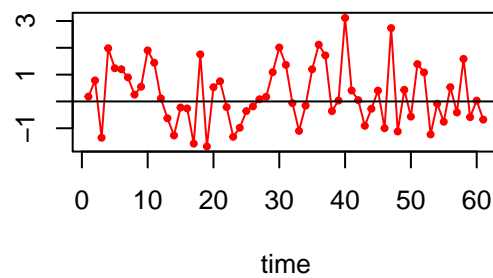
Define a function to produces N AR(1) simulations.

Plot the simulated error terms of the two AR(1)s, that is $u_{x,t}$ and $u_{y,t}$. The respective ACFs indicate that the error terms are not autocorrelated.

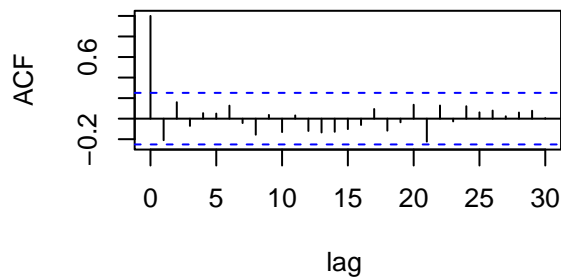
AR innovation: $u_{x,t}$



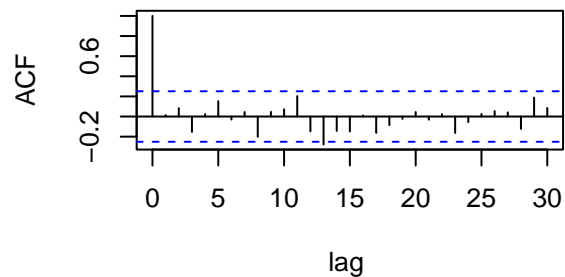
AR innovation: $u_{y,t}$



ACF, AR: $x_t = \phi_1 x_{t-1} + u_{x,t}$



ACF, AR: $y_t = \phi_2 y_{t-1} + u_{y,t}$



Scatterplot of $u_{x,t}$ and $u_{y,t}$. The error terms, while individually and randomly generated, appear to be correlated, which is confirmed by the regression of $u_{y,t}$ on $u_{x,t}$. The resulting t_β is significant.

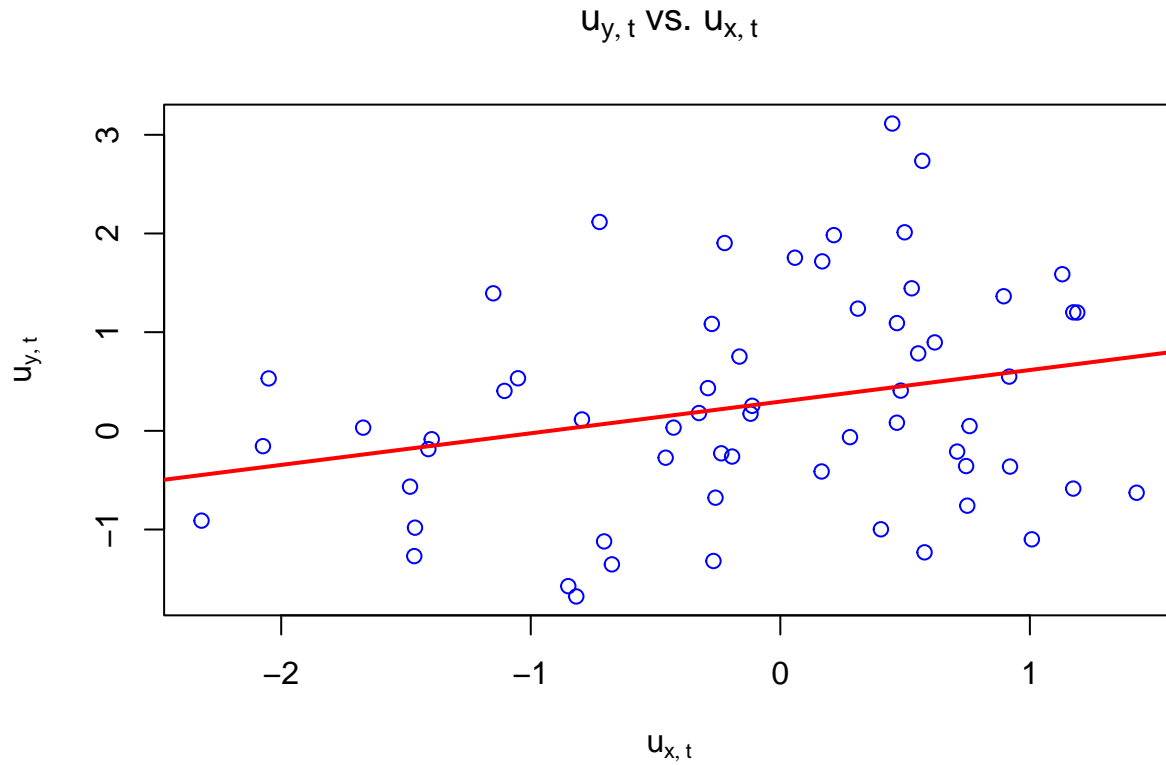
```
par(mfrow=c(1,1))
plot(x = u_sim[,1], y = u_sim[,2],
     col = "blue", lwd = 1,
     xlab=TeX("$u_{x,t}$"), ylab=TeX("$u_{y,t}$"),
     main=TeX('$u_{y,t}$ vs. $u_{x,t}$'))
reg1 <- lm(u_sim[,2] ~ u_sim[,1])
stargazer(list(reg1),type="text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               u_sim[, 2]
## -----
## u_sim[, 1]                    0.321**
##                               (0.151)
##
## Constant                      0.296**
##                               (0.139)
## -----
## Observations                  61
## R2                           0.071
## Adjusted R2                   0.055
## Residual Std. Error          1.080 (df = 59)
## F Statistic                   4.481** (df = 1; 59)
## =====
## Note:                         *p<0.1; **p<0.05; ***p<0.01
```

```
summary(reg1)
```

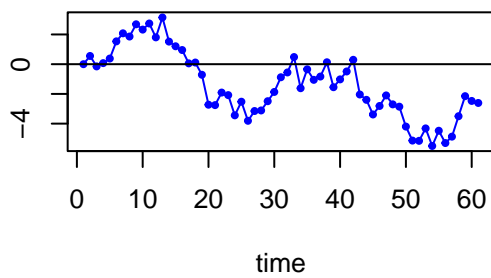
```
##
## Call:
## lm(formula = u_sim[, 2] ~ u_sim[, 1])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.71839 -0.80795 -0.02899  0.64651  2.67481
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.2958     0.1393   2.123  0.0379 *
## u_sim[, 1]    0.3205     0.1514   2.117  0.0385 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.08 on 59 degrees of freedom
## Multiple R-squared:  0.07058,    Adjusted R-squared:  0.05483
## F-statistic: 4.481 on 1 and 59 DF,  p-value: 0.03851
```

```
abline(reg1, col="red", lwd=2)
```

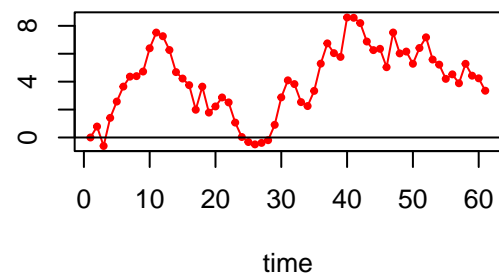


Compare ACF of the two AR processes. Autocorrelation is present and takes a fading wave pattern with significance at the initial 5-10 lags and then again around 15 lags.

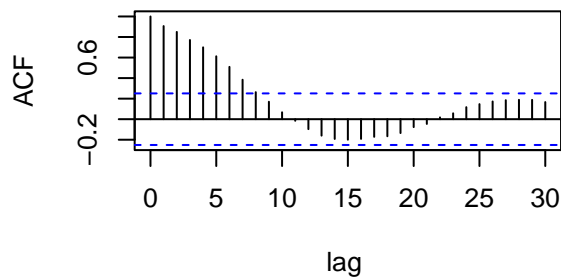
$$\text{AR: } x_t = \phi_1 x_{t-1} + u_{x,t}$$



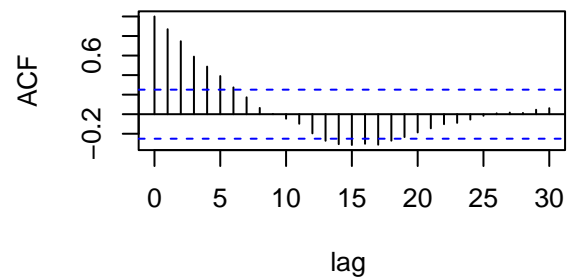
$$\text{AR: } y_t = \phi_2 y_{t-1} + u_{y,t}$$



$$\text{ACF, AR: } x_t = \phi_1 x_{t-1} + u_{x,t}$$



$$\text{ACF, AR: } y_t = \phi_2 y_{t-1} + u_{y,t}$$



Run a regression of two AR(1) processes.

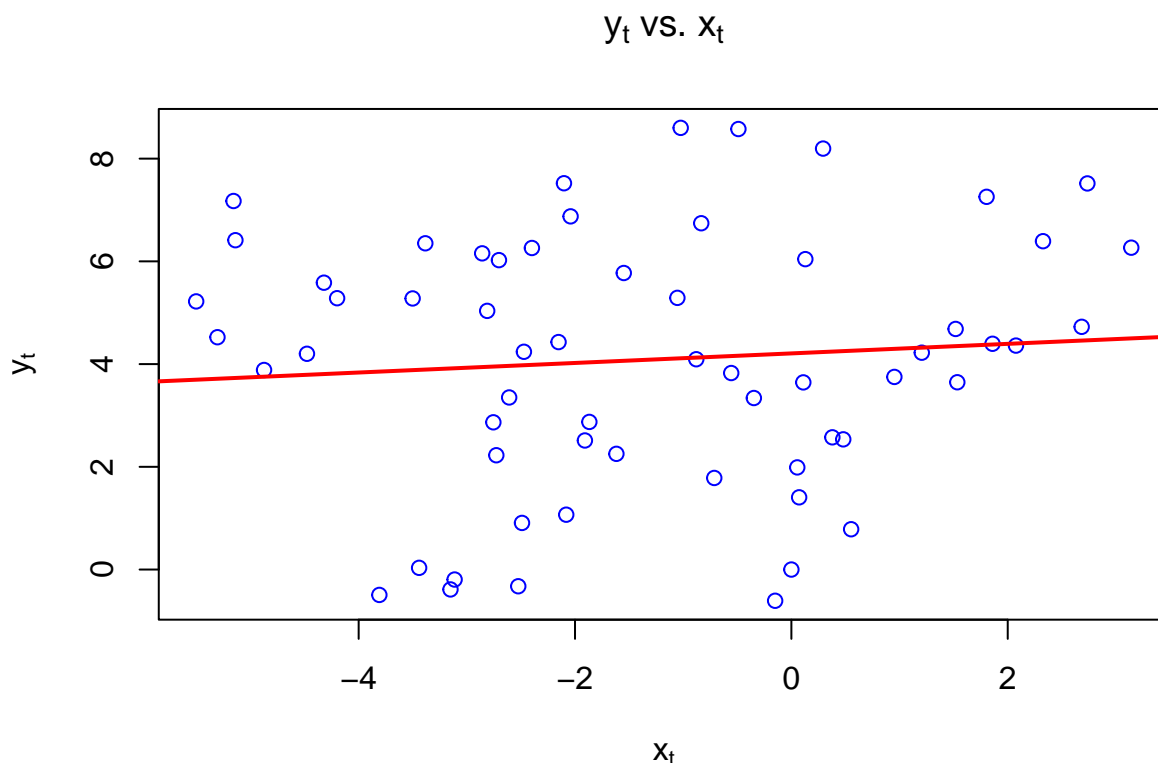
```
#####  
# scatterplot of y_t vs. x_t  
#####  
par(mfrow=c(1,1))  
plot(x = AR_sim[,1], y = AR_sim[,2],  
     col = "blue", lwd = 1,  
     xlab=TeX("$x_{t}$"), ylab=TeX("$y_{t}$"),  
     main=TeX('$y_{t}$ vs. $x_{t}$'), )  
  
reg2 <- lm(AR_sim[,2] ~ AR_sim[,1])  
stargazer(list(reg2),type="text")
```

```
##  
## =====  
##                               Dependent variable:  
##                               -----  
##                               AR_sim[, 2]  
## -----  
## AR_sim[, 1]                    0.093  
##                               (0.143)  
##  
## Constant                      4.208***  
##                               (0.372)  
## -----  
## Observations                  61  
## R2                           0.007  
## Adjusted R2                  -0.010  
## Residual Std. Error          2.479 (df = 59)  
## F Statistic                   0.424 (df = 1; 59)  
## =====  
## Note:                        *p<0.1; **p<0.05; ***p<0.01
```

```
summary(reg2)
```

```
##  
## Call:  
## lm(formula = AR_sim[, 2] ~ AR_sim[, 1])  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -4.8017 -1.7177  0.1307  1.8230  4.4862   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)  4.20802    0.37243   11.299  <2e-16 ***  
## AR_sim[, 1]  0.09304    0.14292    0.651   0.518      
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 2.479 on 59 degrees of freedom  
## Multiple R-squared:  0.007132,    Adjusted R-squared:  -0.009697   
## F-statistic: 0.4238 on 1 and 59 DF,  p-value: 0.5176
```

```
abline(reg2, col="red", lwd=2) # display regression line on scatterplot (y~x)
```



Compare the regression results

```
stargazer(list(reg1,reg2), type = "text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               u_sim[, 2]   AR_sim[, 2]
##                               (1)         (2)
## -----
## u_sim[, 1]                      0.321**
##                               (0.151)
##
## AR_sim[, 1]                      0.093
##                               (0.143)
##
## Constant                      0.296**   4.208***
##                               (0.139)   (0.372)
##
## -----
## Observations                     61       61
## R2                             0.071     0.007
## Adjusted R2                    0.055    -0.010
## Residual Std. Error (df = 59)  1.080     2.479
## F Statistic (df = 1; 59)      4.481**   0.424
## =====
## Note:                          *p<0.1; **p<0.05; ***p<0.01
```

```
# Magics: you can save tabel directly in latex !!!!!!!!!!!!!!!!!!!  
# stargazer(list(reg1,reg2),  
#           align=TRUE, type = "text", no.space = TRUE,  
#           title = "Table X", out = "texoutput/fit.tex")
```