EDHEC PhD Finance 2022 - Econometrics Homework

Giovanni Maffei, Timo Predoehl

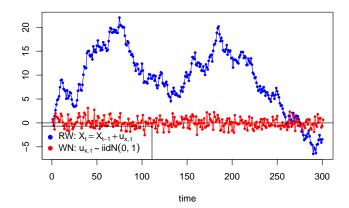
Question 1: Spurious Regressions (2 points)

- 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^2s 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 H0 : $\beta_2 = 0$ 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.

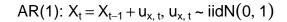
(a)(i) Replication Davidson McKinnon, 2005, fig 14.1

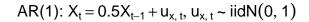
Simulate a random walk processes $y_t = y_{t-1} + e_t, y_0 = 0, e_t \sim IID(0, \sigma^2)$ analog to (14.03)

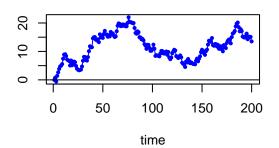
White Noise and Random Walk

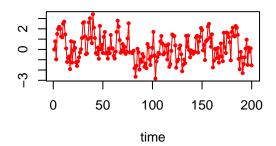


Simulate two AR(1) processes according to $x_t = \phi x_{t-1} + u_t$, with $\phi = 0.5$ and with $\phi = 1$. The ACFs of two processes indicate that in latter significant autocorrelation persists at least up to 20 lags, while the former autocorrelation becomes insignificant after 3 lags. Hence, in the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



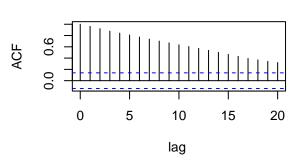


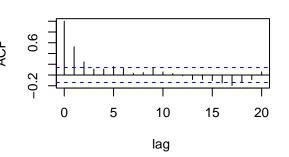




ACF, AR(1) NON-stationary

ACF, AR(1) stationary



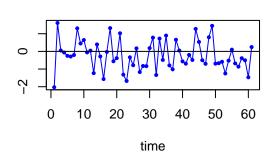


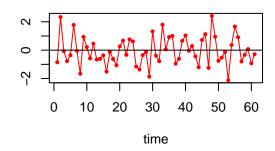
Define a function to produces N AR(1) simulations.

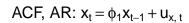
Plot the simulated error terms of the two AR(1)s, that is $u_{x,t}$ and $u_{y,t}$. The respective ACFs indicate that the error terms are not autocorrelated.

AR innovation: ux.t

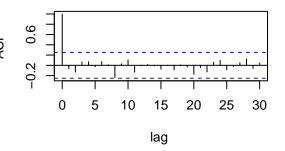
AR innovation: u_{y, t}

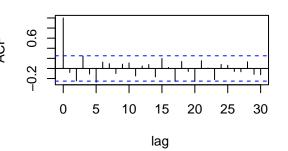






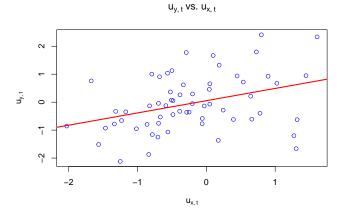
ACF, AR: $y_t = \phi_2 y_{t-1} + u_{v,t}$





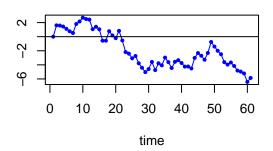
Scatterplot of $u_{x,t}$ and $u_{y,t}$. The error terms, while individually and randomly generated, appear to to correlated, which is confirmed by the regression of $u_{y,t}$ on $u_{x,t}$. The resulting t_{β} is significant.

```
##
                      Dependent variable:
##
                   -----
##
                         u_sim[, 2]
##
  ______
## u_sim[, 1]
                          0.438***
##
                           (0.147)
##
## Constant
                            0.053
##
                           (0.125)
##
##
## Observations
                             61
## R2
                            0.131
## Adjusted R2
                            0.116
## Residual Std. Error
                      0.943 \text{ (df = 59)}
## F Statistic
                     8.895*** (df = 1; 59)
## Note:
                   *p<0.1; **p<0.05; ***p<0.01
##
## Call:
## lm(formula = u_sim[, 2] ~ u_sim[, 1])
## Residuals:
##
             1Q Median
                          ЗQ
     Min
                                Max
## -2.2890 -0.5590 -0.0190 0.4804 2.0071
##
## Coefficients:
            Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.05336
                      0.12516
                             0.426 0.67143
## u_sim[, 1]
            0.43838
                      0.14699
                              2.982 0.00415 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9435 on 59 degrees of freedom
## Multiple R-squared: 0.131, Adjusted R-squared: 0.1163
## F-statistic: 8.895 on 1 and 59 DF, p-value: 0.004152
```

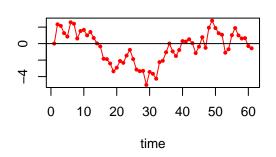


Compare ACF of the two AR processes. Autocorrelation is present and takes a fading wave pattern with significance at the initial 5-10 lags and then again around 15 lags.

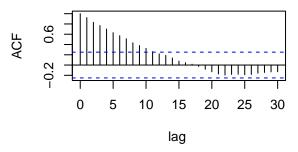
AR:
$$x_t = \phi_1 x_{t-1} + u_{x, t}$$



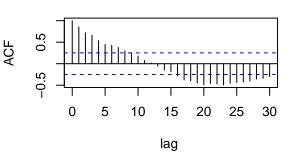
AR:
$$y_t = \phi_2 y_{t-1} + u_{v,t}$$



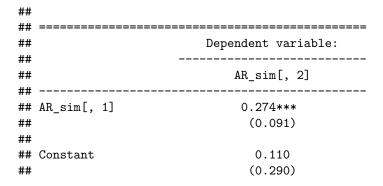
ACF, AR:
$$x_t = \phi_1 x_{t-1} + u_{x, t}$$



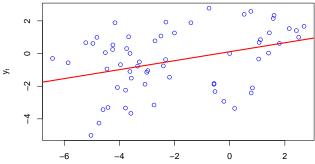
ACF, AR:
$$y_t = \phi_2 y_{t-1} + u_{y, t}$$



Run a regression of two AR(1) processes.



```
##
## -----
## Observations
                             0.132
## R2
## Adjusted R2
                             0.118
## Residual Std. Error
                         1.777 \text{ (df = 59)}
## F Statistic
                      8.997*** (df = 1; 59)
*p<0.1; **p<0.05; ***p<0.01
## Note:
##
## Call:
## lm(formula = AR_sim[, 2] ~ AR_sim[, 1])
##
## Residuals:
             1Q Median
##
      Min
                            ЗQ
## -3.7395 -1.1474 0.2007 1.4091 2.9137
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.11036
                       0.29011
                                 0.38 0.70500
                                 3.00 0.00396 **
## AR_sim[, 1] 0.27438
                       0.09148
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.777 on 59 degrees of freedom
## Multiple R-squared: 0.1323, Adjusted R-squared: 0.1176
## F-statistic: 8.997 on 1 and 59 DF, p-value: 0.003956
```



 \mathbf{x}_{t}

Compare the regression results using stargazer:

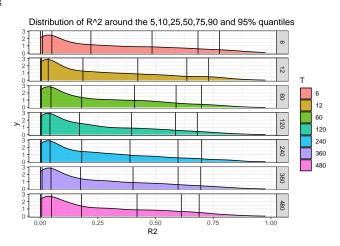
```
##
 ______
##
                           Dependent variable:
##
##
                         u_sim[, 2]
                                    AR_sim[, 2]
                                    (2)
##
## u_sim[, 1]
                          0.438***
##
                          (0.147)
##
## AR_sim[, 1]
                                    0.274***
```

## ##			(0.091)
	Constant	0.053	0.110
##		(0.125)	(0.290)
##			
##			
##	Observations	61	61
##	R2	0.131	0.132
##	Adjusted R2	0.116	0.118
##	Residual Std. Error (df = 59)	0.943	1.777
##	F Statistic (df = 1; 59)	8.895***	8.997***
##		=========	=========
##	Note:	*p<0.1; **p<0.05; ***p<0.01	

Now lets do the 1,000,000 simulations of 2 AR(1)s, regress them on each other and store the regression results in a matrix for further analysis.

23.417 sec elapsed

Plot histogram of results



The rate of rejection of $H_0: \beta = 0$ increases with the sample size T.

