

## Master 2 M.M.M.E.F. 2016 – 2017

Time Series Tutorial  $n^0$  2 :

## How to identify a white noise and generate ARMA and GARCH processes?

The aims of this tutorial is a to provide a first tool to identify a white noise and to generate both the classical ARMA and GARCH processes.

## Identification of a white noise: test of portemanteau

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 Vizualization of the independance
 

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```
x=rnorm(50)
r=acf(x)
```

Generate a Gaussian white noise.  
Correlogram of  $x$

It represents empirical auto-correlations for consecutive lags  $0, 1, 2, \dots$ . Dots represent 95% confidence intervals of  $Z/\sqrt{n}$ , where  $Z$  is  $\mathcal{N}(0, 1)$  r.v. and  $n$  is the length of  $x$ .

Why? Which test is associated to the confidence interval? Can we consider  $x$  as a white noise?

```
y=x[1:49]+x[2:50]
```

Generation of a new trajectory. Which kind of process is  $y$ ?

What is the distribution of  $y$ ? Compute the correlations of  $y$  for any lag.

```
ry=acf(x)
rho=ry$acf[1]
```

Correlogram of  $y$ . Is it a white noise?

Computation of the empirical correlation for lag 1.

Compare with the theoretical correlation.

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 Goodness-of-fit and whiteness tests
 

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```
x=5*rnorm(50)-2
hist(x,nclass=11)
```

What is the law of  $x$ ?

Histogram which is a first estimation of the density of  $x$ . Why?

Note the choice of numbers of classes of histogram.

```
qqnorm(x)
y=3*runif(100)+1
```

QQplot test (what is it?). Conclusion?

what is the law of  $y$ ?

```
qqnorm(y)
dn=ks.test(x,"pnorm",-2,5)
```

Conclusion?

Kolmogorov-Smirnov test where the distribution of  $x$  is compared to  $\mathcal{N}(-2, 5^2)$ .

The  $p$  – value provides a quantitative way for deciding from a test.

Classically  $H_0$  is accepted when  $p$  – value  $> 0.05$ .

```
ddn=ks.test(y,"pexp",3)
```

Test on  $y$ . What could we expect?

Test now if  $y$  follow a uniform law on  $[-1, 1]$ .

```
ks.test(x,y)
```

Test the similarity of the distributions of  $x$  and  $y$ . Conclusion?

## ARMA process

In the sequel, two different ways are followed for generating a trajectory of a ARMA process.

```
X=arima.sim(100,model=list(ar=-.3,ma=.7))
```

Simulation of a trajectory of ARMA[1,1] process.

Write the recurrent equation followed by this process.

Draw this trajectory.

Representation of its correlogram. Conclusion?

Generate another trajectory using directly the recurrent equation.

Generate a trajectory of a ARMA[2,2] process (chosed the coefficients).

Generate the same trajectory with a noise following a uniform distribution on  $[-1, 1]$ .

We generate another trajectory using directly the recurrent equation.

```
n=100; m=100;
epsi=3*rnorm(n+m)
X=0
for (j in c(1:(n+m)))
  X[j+1]=-0.3*X[j]+epsi[j+1]+0.7*epsi[j]
x=X[c((m+1):(n+m+1))]; tsplot(x)    Explain what is done here.
```

## GARCH process

Let  $X$  be a trajectory of length 100 of the following process:

$$X_t = \varepsilon_t \times \sigma_t \quad \text{and} \quad \sigma_t^2 = 4 + 0.3 \times X_{t-1}^2 + 0.4 \times \sigma_{t-1}^2,$$

where  $(\varepsilon_t)$  is a Gaussian standard white noise.

Generate directly this trajectory (as previously for ARMA process let run the routine  $m = 100$  before for being close to a stationary process). Draw this trajectory and the correlogram.

Do the same thing by using the R package `fGarch`.