

Practice 2 R: Simulation of time series

A) IID and Gaussian white noises

Do

- `n<-1000`
- `z1<-rnorm(n)`
- `z1`

We have simulated a sample of n data with standard normal law. In Time Series language, this is a standard Gaussian White Noise. Note that we can do directly

- `z1<-rnorm(100)`
- `z1`

but the first option gives more programming flexibility.

If we do

- `mu=5`
- `sigma=2`
- `z2<-rnorm(n,mean=mu,sd=sigma)`
- `z2`

we simulate a sample of a normal random variable with mean 5 and standard deviation 2 or in other words a Gaussian White Noise with mean 5 and standard deviation 2.

Try also

- `m<-1`
- `set.seed(m)`
- `w<-rnorm(1000)`
- `w`

and choose different values for the value of m , different than 1.

Note that by default R choose a random seed. If we fix the seed, we obtain the same sample repeatedly.

We can plot the graphics of our samples using for example

➤ `plot(z2,type="l")`

Of course, you can improve the plot using instructions as `xlab`, `ylab` and so on. See the help of the instruction `plot`.

If we do

➤ `hist(z1,freq=F)`

we obtain the histogram of a standard normal sample. See the help of the instruction `hist`. What happens if we erase `freq=F`?

Repeat the same exercise for different values of `n`, `mu` and `sigma` and observe what happens.

To compare how samples differs from their underlying population do

➤ `n<-1000`
➤ `x<-seq(-4,4, length=1000)`
➤ `hist(rnorm(n), prob=T); points(x,dnorm(x), type="l")`

Try `n` greater.

It is important also to compute the correlogram.

➤ `N=10000`
➤ `mu=5`
➤ `sigma=2`
➤ `z<-rnorm(n,mean=mu,sd=sigma)`
➤ `z`
➤ `acf(z)`

Observe that it is a correlogram of an IID-noise.

B) Random walk:

With instructions

- `n<-1000`
- `z<-rnorm(n)`
- `z`
- `x<-cumsum(z)`
- `plot(x,type="o")`

we obtain a trajectory of a standard random walk. Do the same for other Gaussian white noises with mean and variance.

Another way to do it is the following:

- `n<-10000`
- `w<-rnorm(n)`
- `x<-w`
- `for (t in 2:n) x[t]<-x[t-1]+w[t]`
- `plot(x,type="l")`

Finally, we can see that first differences are an IID noise:

- `acf(diff(x))`