

White Noise and Random Walk

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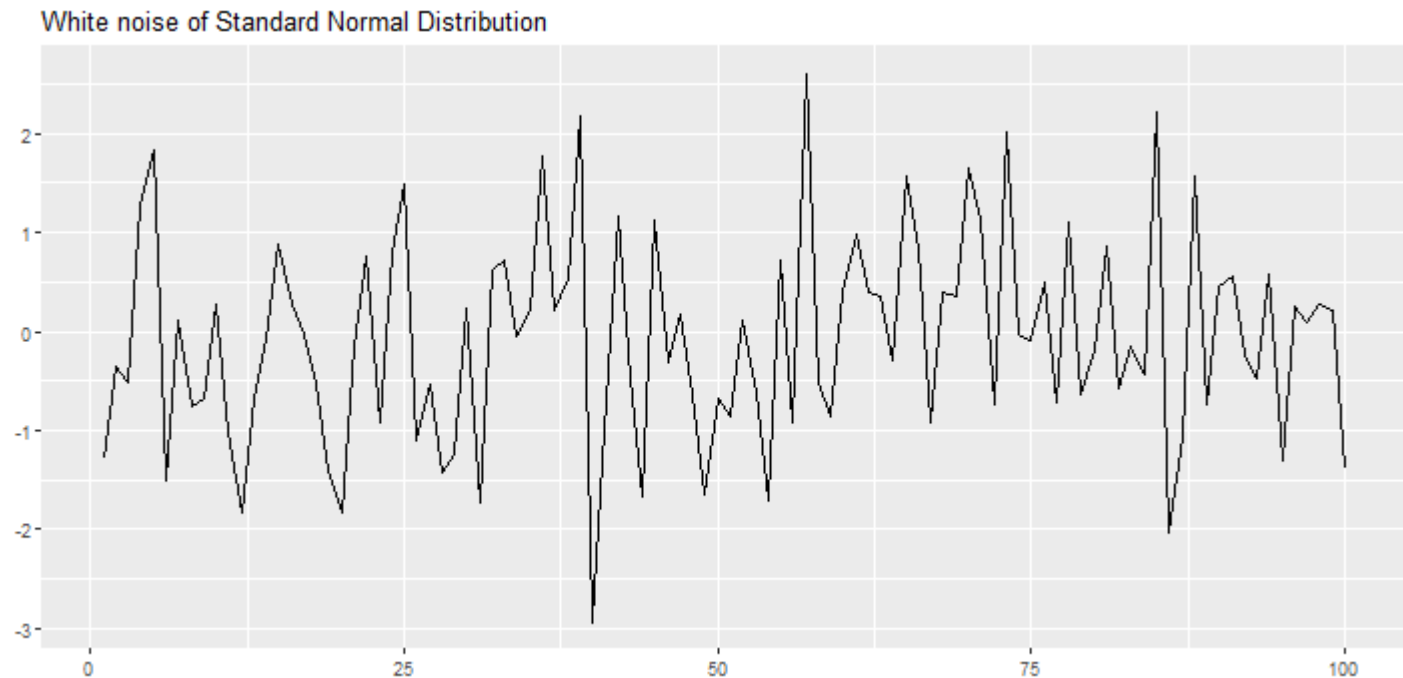
White Noise

- y_t is a white-noise process (series) if y_1, y_2, \dots, y_t are independent identical distributed (iid) zero mean random variables from a certain distribution (usually normal)

Example

noise from normal dist.

```
set.seed(30)
y <- ts(rnorm(100))
library(ggfortify)
autoplot(y) + ggtitle("White noise of Standard Normal Distribution")
```



Example

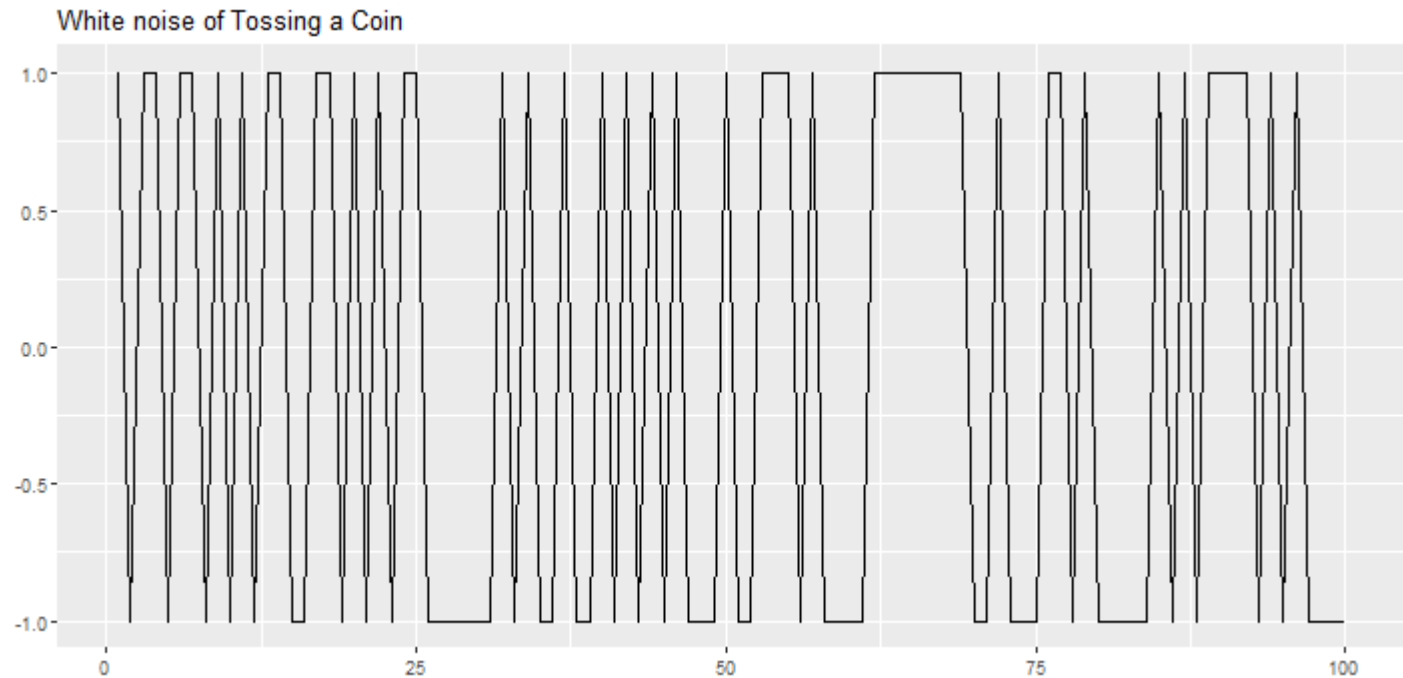
```
set.seed(30)
```

```
y = sample(c(-1, 1), 100, replace = TRUE)
```

```
y ← ts(y)
```

```
library(ggfortify)
```

```
autoplot(y) + ggtitle("White noise of Tossing a Coin")
```



Correlogram

- Autocorrelation lag with lag k is the correlation between the time series y_t and y_{t-k}

$$\rho_k = \frac{\sum_{t=k+1}^n (Y_t - \bar{Y})(Y_{t-k} - \bar{Y})}{\sum_{t=1}^n (Y_t - \bar{Y})^2}$$

where:

- ρ_k : Autocorrelation at lag k
- Y_t : Value of the series at time t
- \bar{Y} : Mean of the series
- n : Number of observations

- Autocorrelation lag with lag 0 is always 1

- The Correlogram is the plot of the autocorrelations for values of lag $k = 0, 1, 2, \dots$

correlation between x_t & x_{t-k} measures the linear relation

auto correlation \sim correlation between (x_t, x_{t-1})

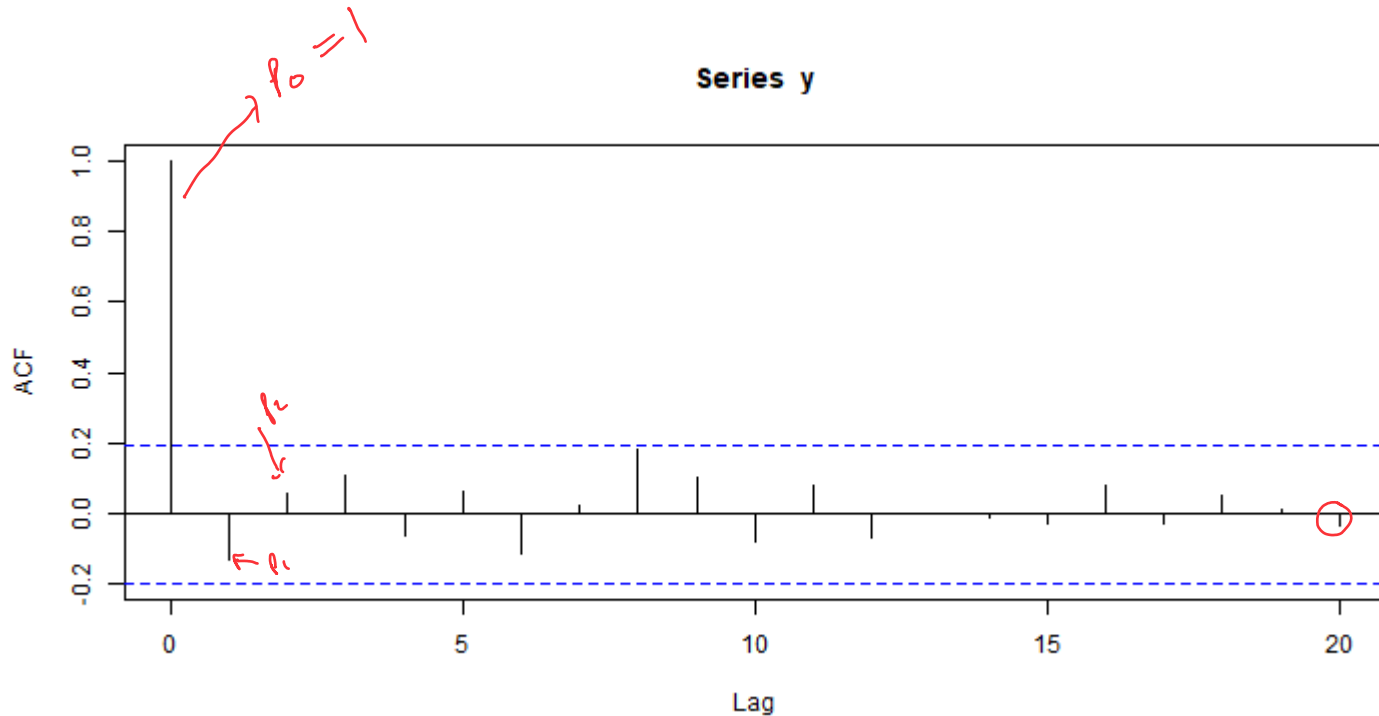
lag 1

lag k : (x_t, x_{t-k})

Correlogram a white noise

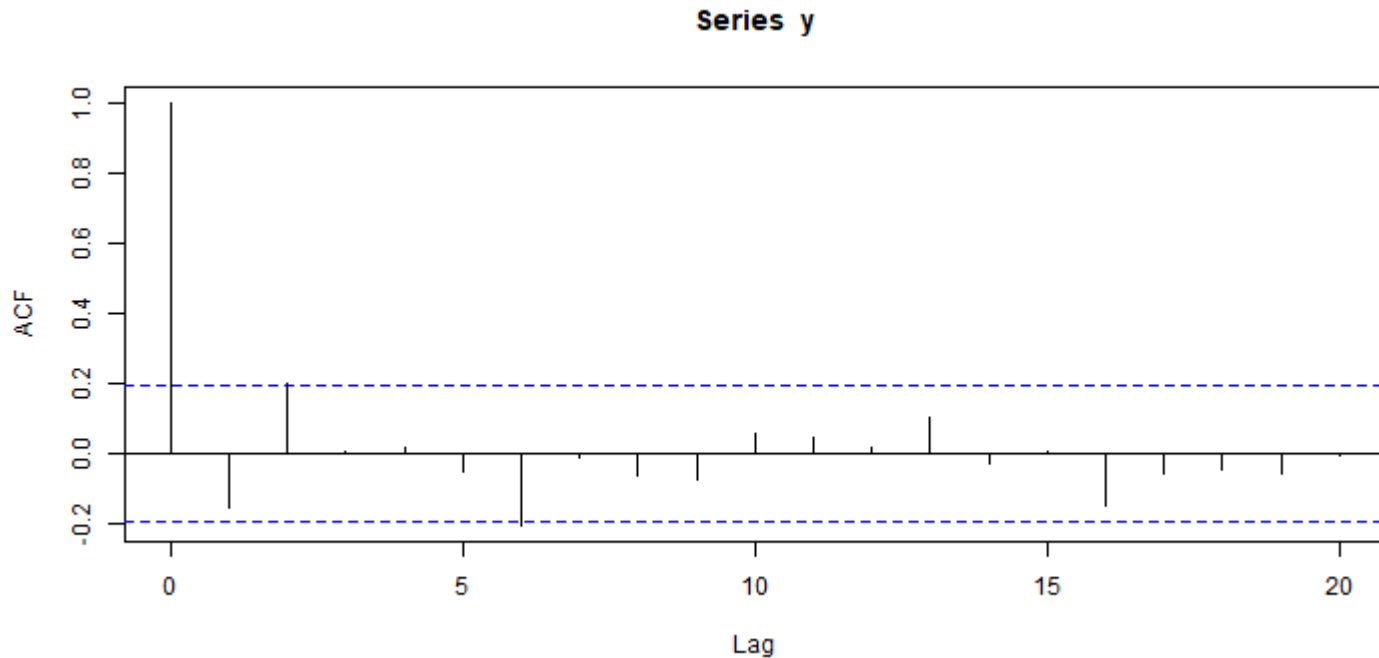
- Correlogram of a white noise

```
# create a white-noise time series  
y = ts(rnorm(100))  
  
# plot its ACF or correlogram  
acf(y)
```



Correlogram a white noise

```
set.seed(30)
y = sample(c(-1, 1), 100, replace = TRUE)
y <- ts(y)
acf(y)
```

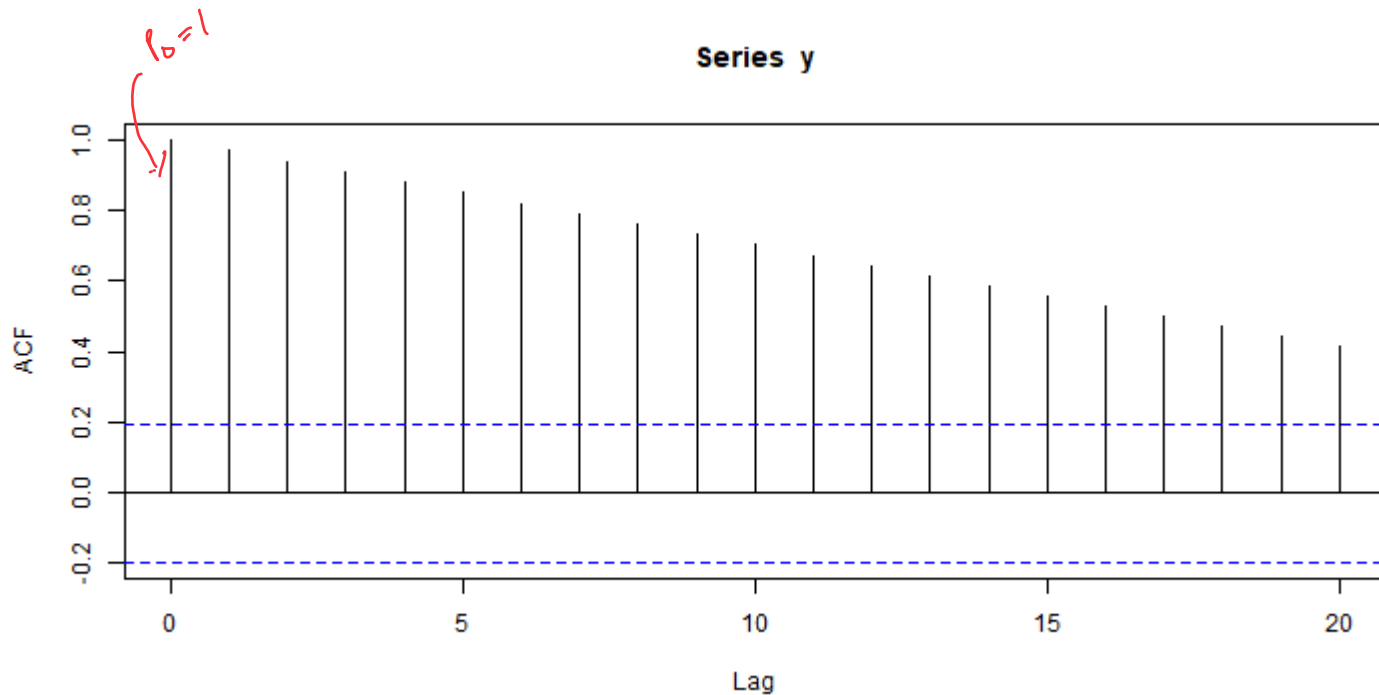


Correlogram a time series with trend

- Usually a trend in the data will show in the correlogram as a slow decay in the autocorrelation

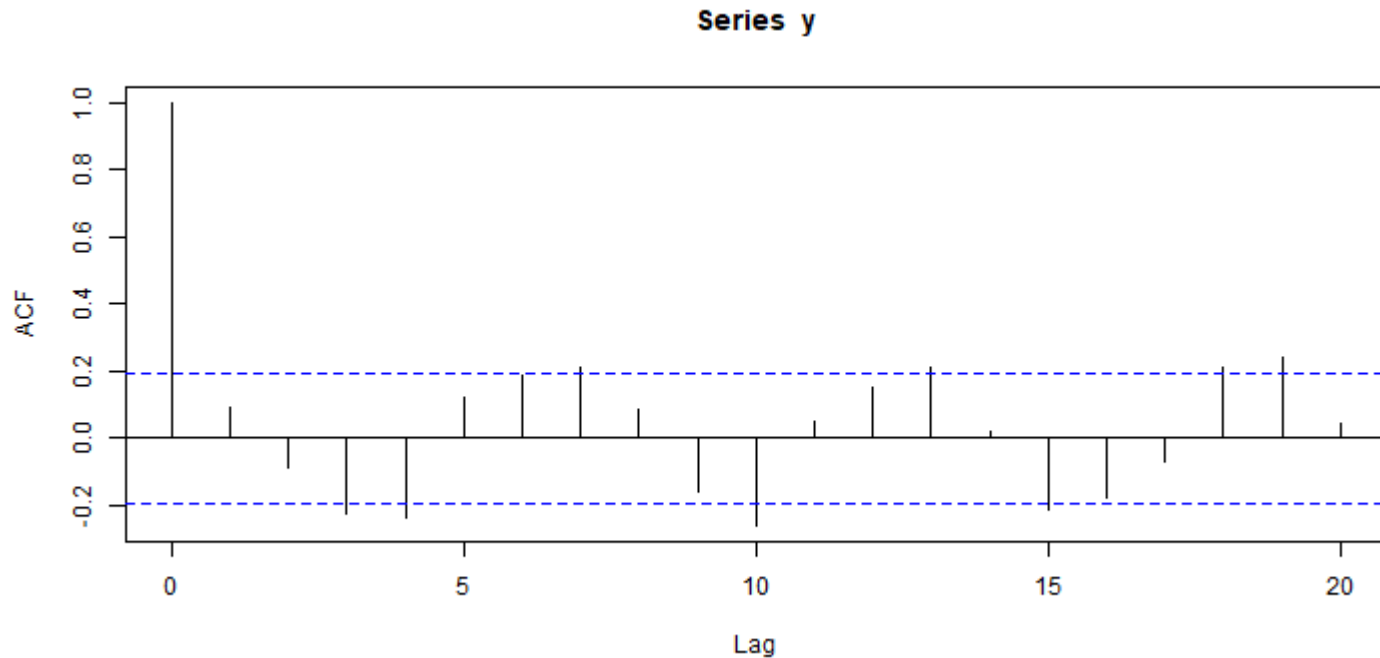
$$y_t = \{1, 2, 3, \dots, 100\}$$

```
y = ts(c(1:100))  
acf(y)
```



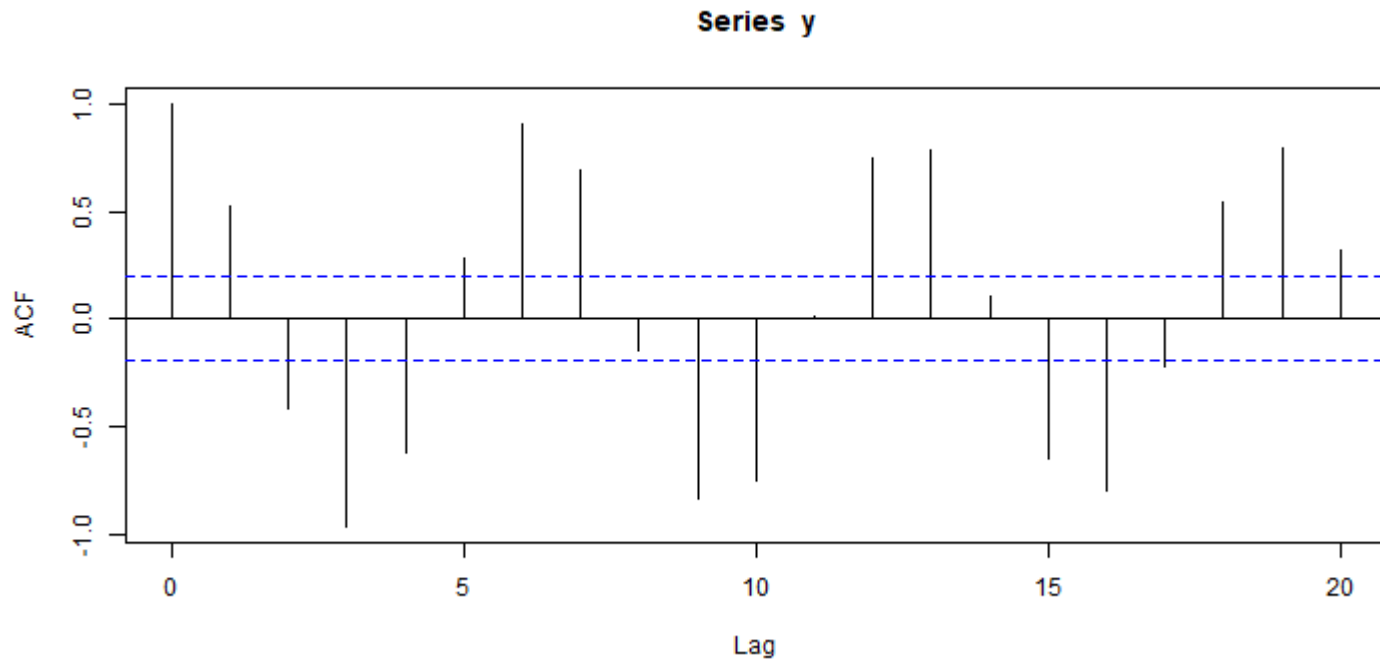
The Correlogram - Example

```
y = ts(cos(c(1:100))+rnorm(100))  
acf(y)
```



ACF of a time series with seasonality

```
set.seed(30)
y = cos(1:100)
y <- ts(y)
acf(y)
```



Random Walk

- A time series y_t is called a random walk if

$$y_t = y_{t-1} + \epsilon_t,$$

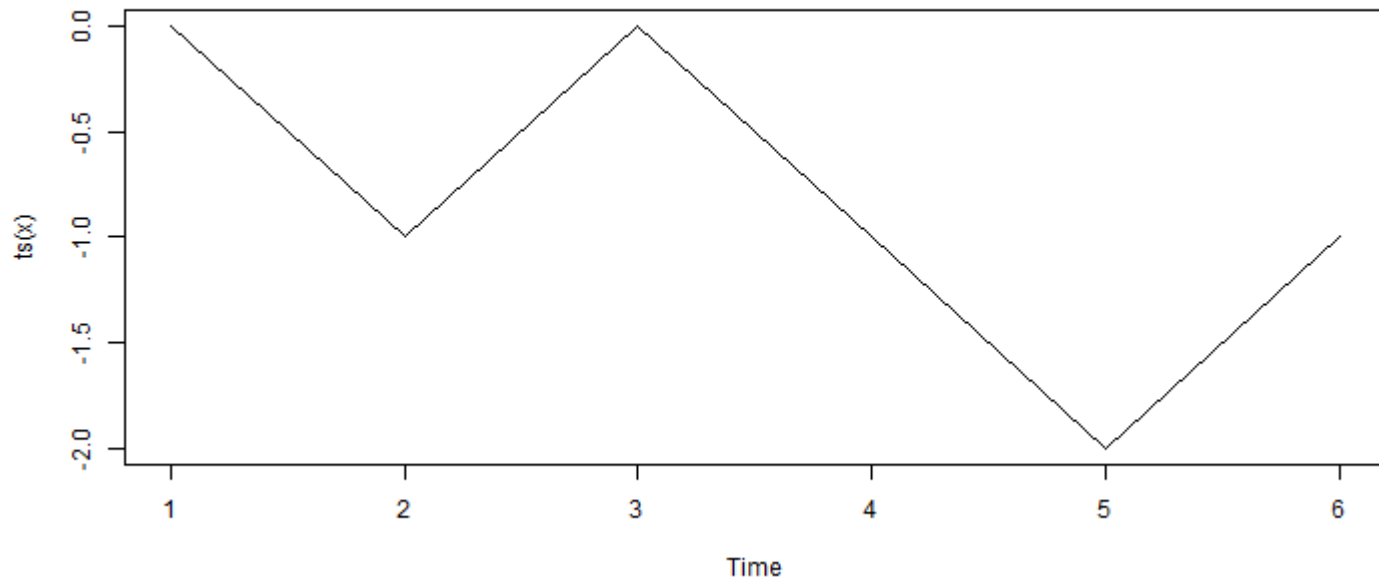
where ϵ_t is a white-noise

- A random walk can be written as

$$y_t = y_0 + \epsilon_1 + \epsilon_2 + \dots + \epsilon_t$$

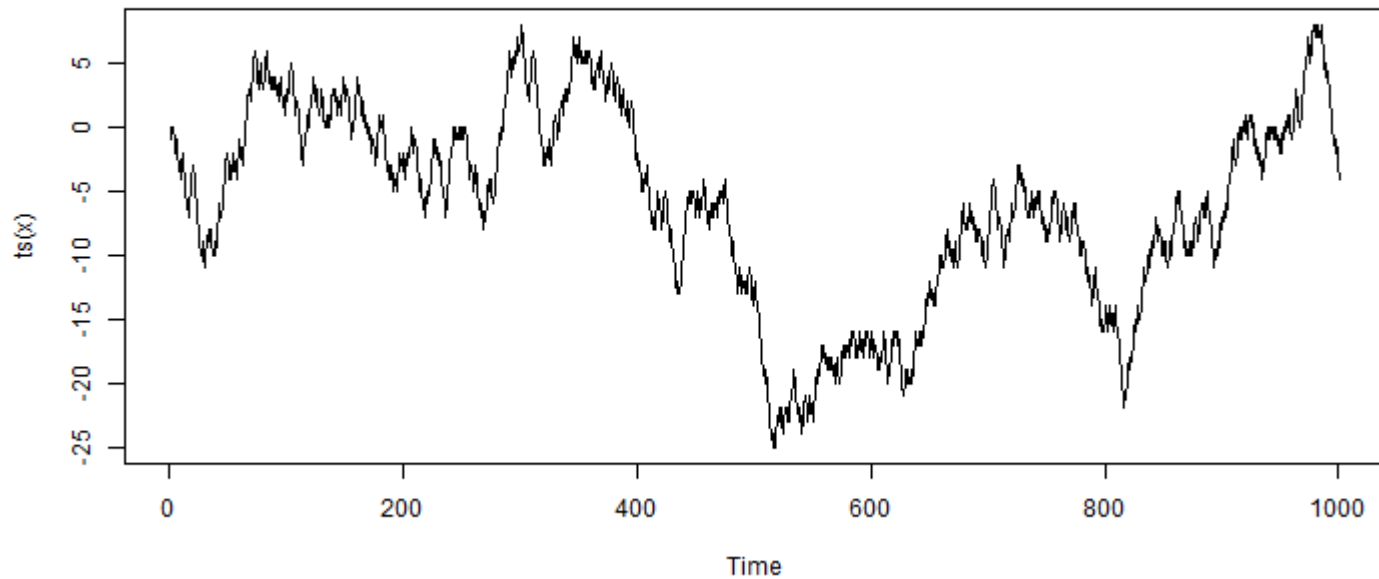
Example

```
set.seed(1)
n <- 5
ct = sample(c(-1, 1), n, TRUE)
x <- cumsum(c(0,ct))
plot(ts(x))
```



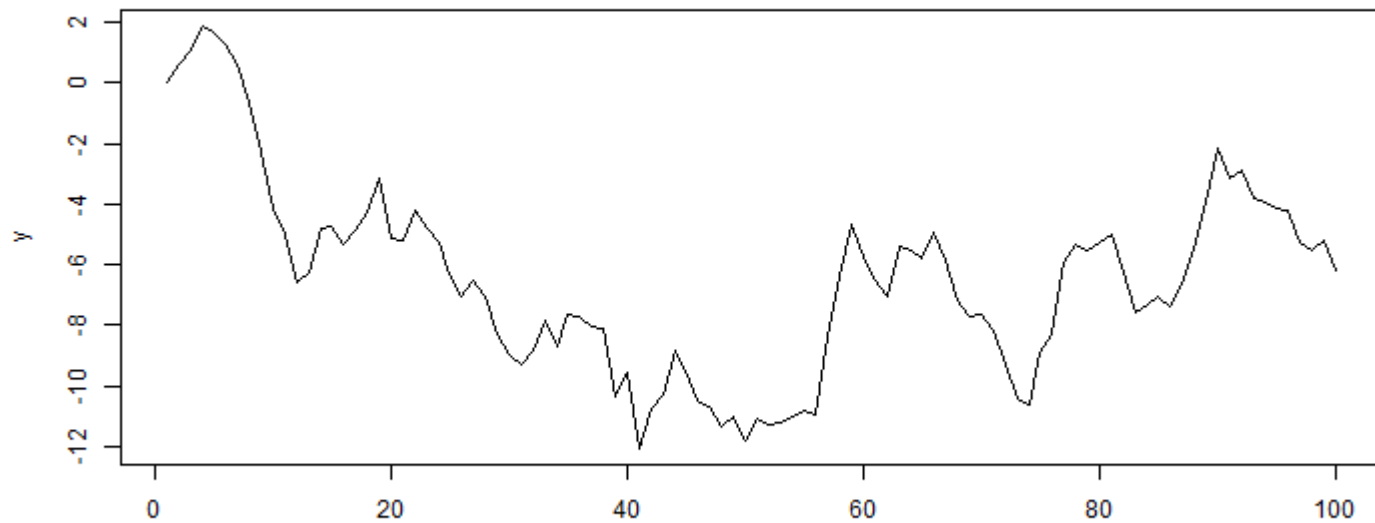
Example

```
set.seed(1)
n <- 1000
ct = c(0, sample(c(-1, 1), n, TRUE))
x <- cumsum(ct)
plot(ts(x))
```



Example

```
set.seed(3000)
n = 100
c ← rnorm(n)
y_0 = 0
y = c(y_0, 2:n)
for (i in 2:n)
{
  y[i] = y[i-1]+c[i]
}
y = ts(y)
plot(y)
```



Random Walk with drift

- A time series y_t is called a random walk if

$$y_t = y_{t-1} + d + \epsilon_t,$$

where ϵ_t is a white-noise

- A random walk can be written as

$$y_t = y_0 + dt + \epsilon_1 + \epsilon_2 + \dots + \epsilon_t$$

Example

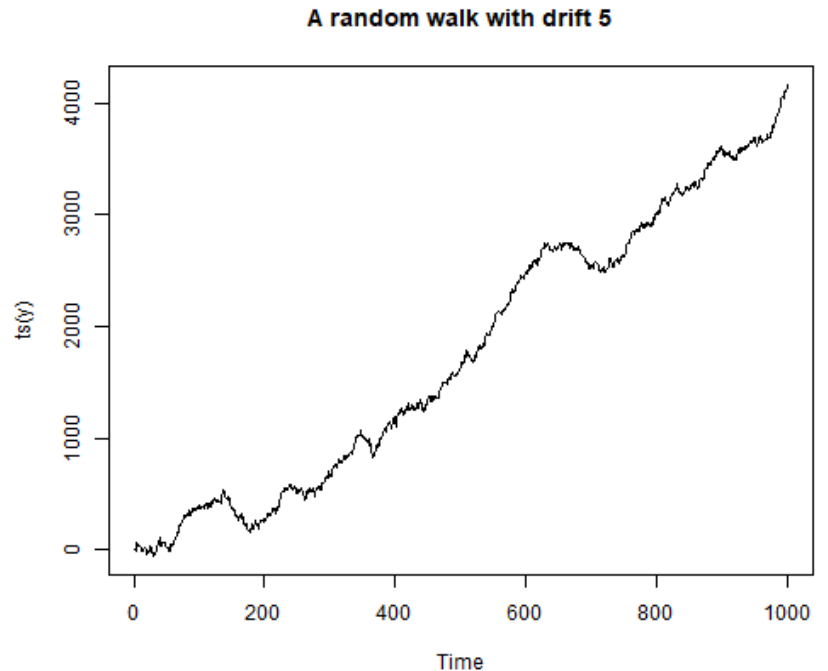
```
set.seed(30)
n = 1000
c ← rnorm(n, sd = 20)
y_0 = 0
drift = 5

y = c(y_0, 2:n)

for (i in 2:n)
{
  y[i] = drift + y[i-1] + c[i]
}

library(ggfortify)
library(latex2exp)

plot(ts(y))
title(paste0("A random walk with drift ", drift))
```

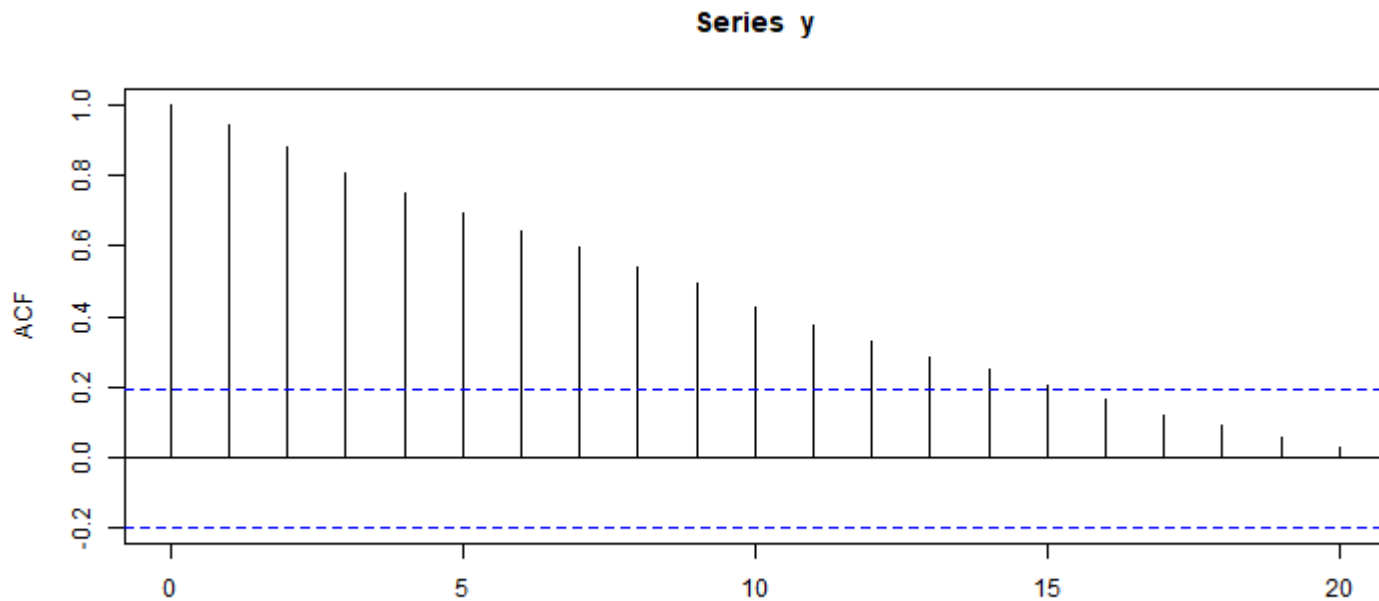


The ACF of Random Walks

```
n = 100
error_mean = 0
c ← rnorm(n, mean = error_mean, sd = 30)
y_0 = 0
y = c(y_0, 2:n)

for (i in 2:n)
{
  y[i] = y[i-1]+c[i]
}

acf(y)
```

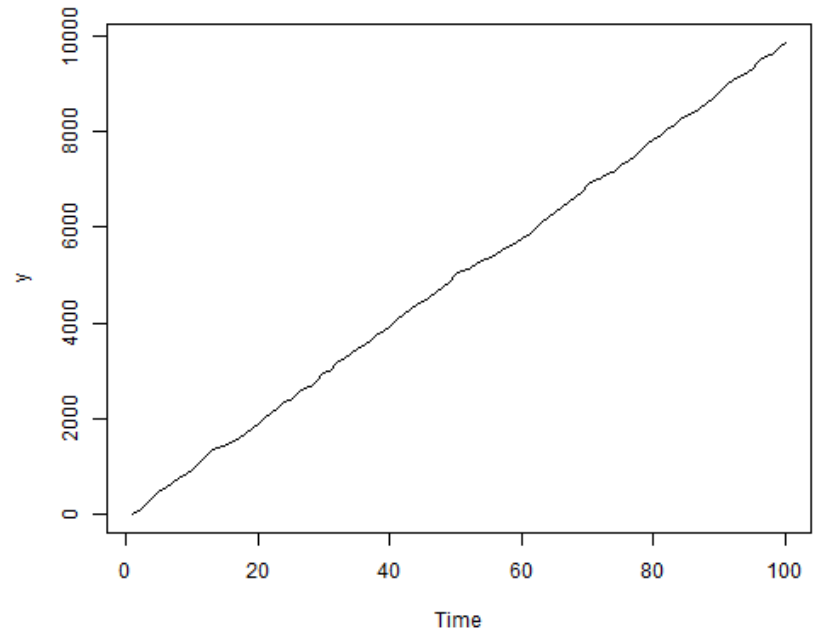


Differencing Time Series

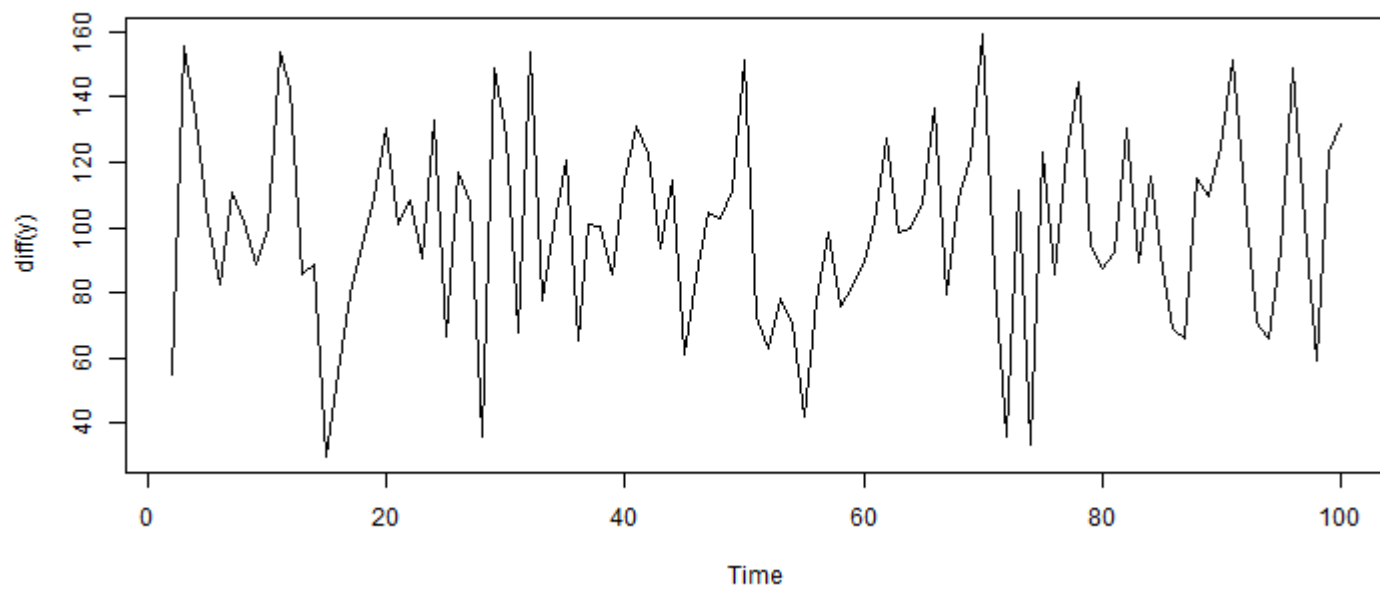
```
n = 100
error_mean = 0
drift = 100
c <- rnorm(n, mean = error_mean, sd = 30)
y_0 = 0
y = c(y_0, 2:n)

for (i in 2:n)
{
  y[i] = drift + y[i-1]+c[i]
}

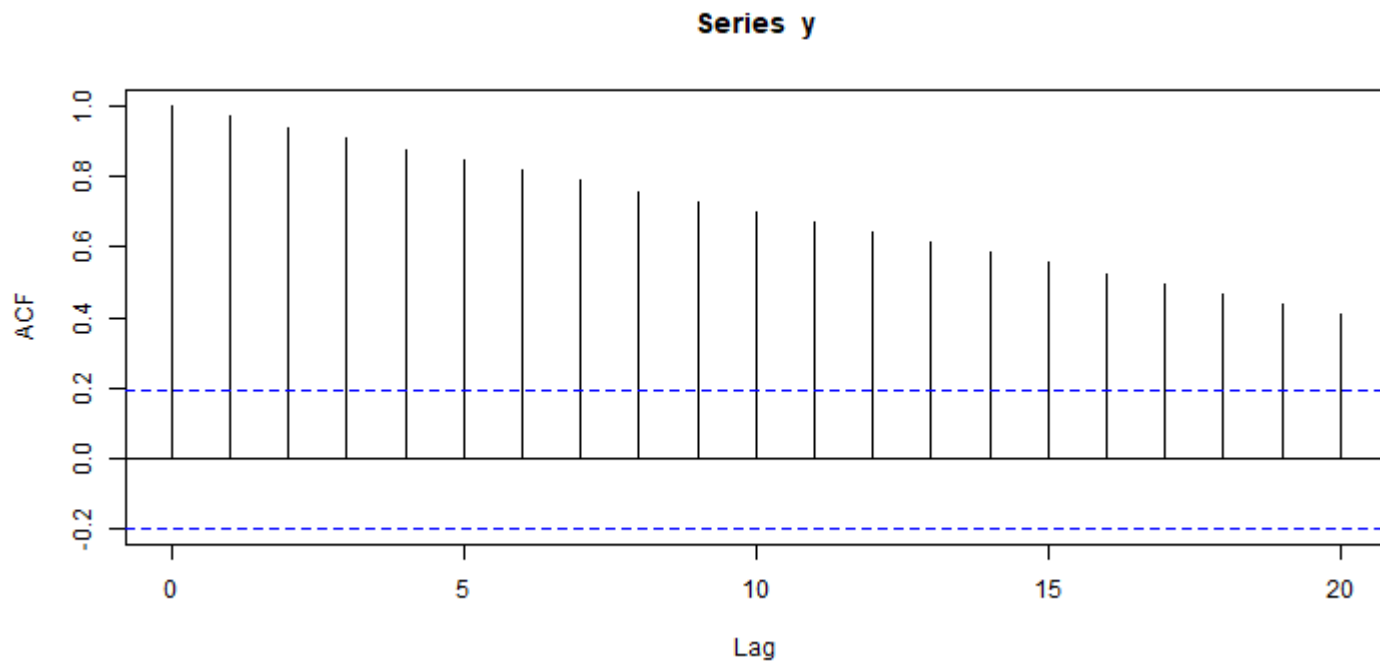
y = ts(y)
plot(y)
```



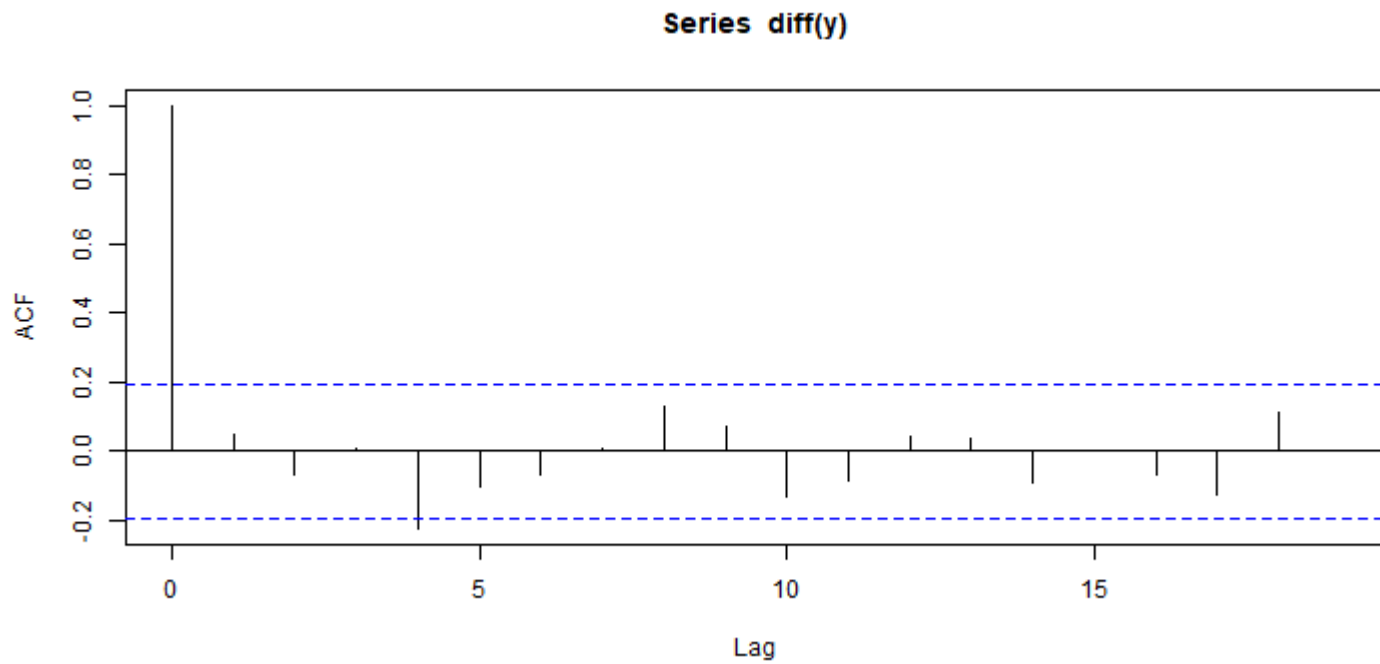
```
plot(diff(y))
```



acf(y)



```
acf(diff(y))
```



Quantmod Package

<https://www.quantmod.com/>

Random Walks and Stocks

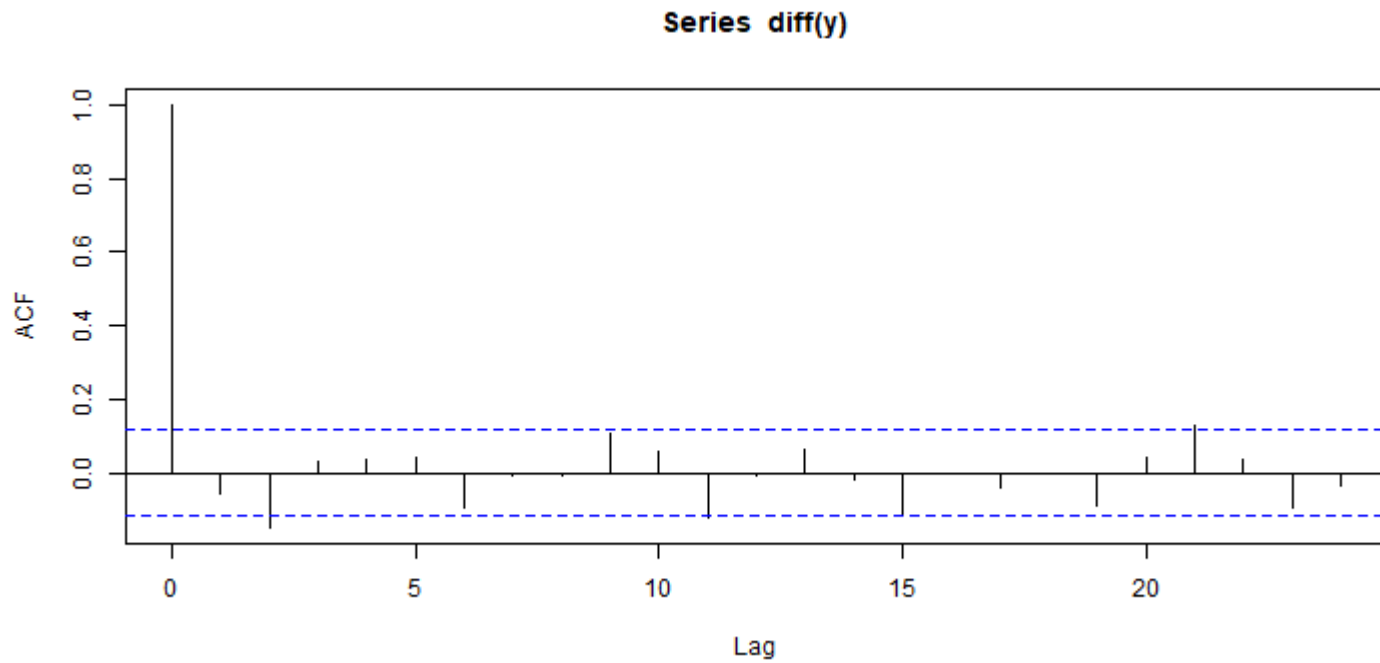
```
library(quantmod)
getSymbols('MSFT', src='yahoo')
```

```
## [1] "MSFT"
```

```
y = Ad(MSFT[index(MSFT)>"2023-01-01",])
plot(y)
```



```
acf(diff(y), na.action = na.pass)
```



- The differencing series could be a white noise
- It is very reasonable to assume that the stock follows the random walk model.