

```
## Registering fonts with R
## Package "fontcm" already installed.
## Registering font package "fontcm" with fonts.
## Font package "fontcm" already registered in fonts database.
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:plyr':
##
##   arrange, count, desc, failwith, id, mutate, rename, summarise,
##   summarize
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

1 Description

The symmetric random walk will be described in this document (Mt). it covers the theory of "Stochastic Calculus for finance" Tome 2 chapter 3 section 1.

The construction of the random walk depend on the evolution of a random variable X_i . The previous RV can take two value at each time, like tossing a coin. X_i can take the value 1 or -1.

$$X_i = \begin{cases} 1 \\ -1 \end{cases} \quad (1)$$

The Symetric Random Walk is constructed by summing up the different outcome of the random variable X_i from k experiments:

$$M_k = \sum_{j=1}^k X_j \quad (2)$$

In the following lines of code, X_i is randomly difined. The variable k ensure to have a sufficient number of periods to further generate the scaled random walk. It refers to the k of equation 2. p and q are the probability measure, respectively p chance to get value 1 and q chance to get -1 from random variable X_i .

After creating the random variable X_i it suffices to add up all the differente output we get from time 1 up to k to get a specific Symetric Random Walk.

The following outcome present a randomly generated 300 steps symmetric random walk.

Table 1: 300 steps Symmetric Random Walk

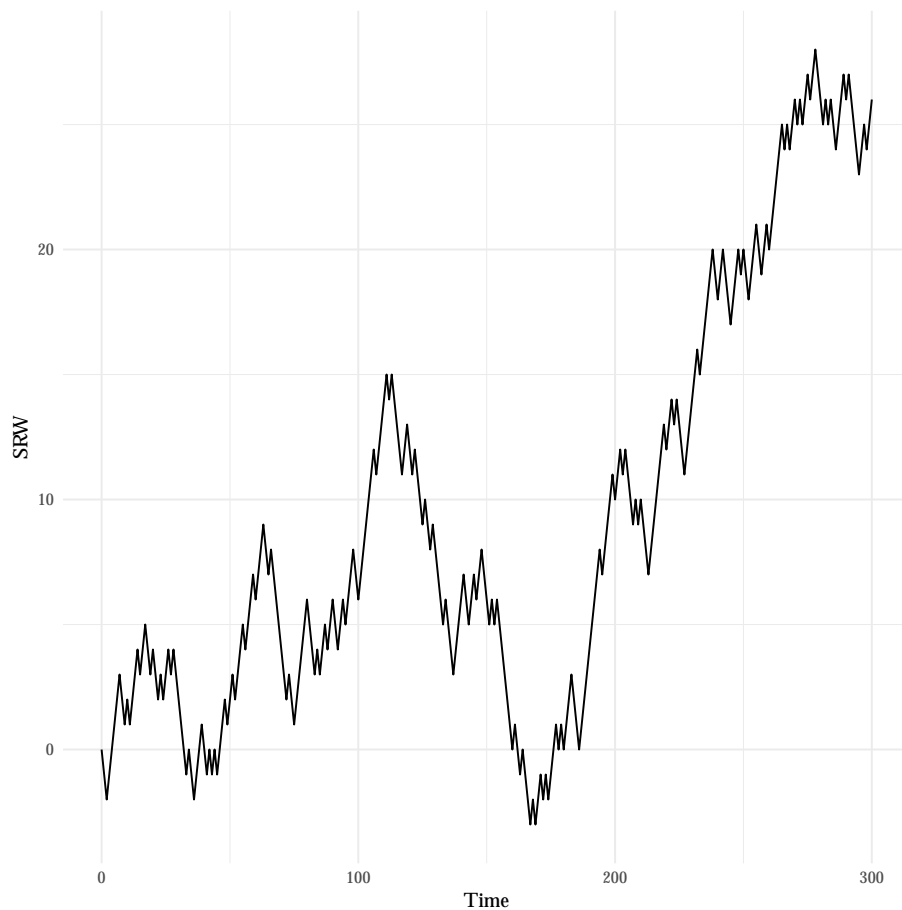


Figure 1: Symmetric Random Walk

```
# Because squared matrix dim(y) = dim(x):
dim_x <- dim_y <- 1:(k + 1) # from 1 to k+1 because we start to time zero nonrandom which e
o <- outer(dim_x, dim_y, FUN=function(r,c){(r-c) + (1-c)} )
to <- t(o)
subset <- upper.tri(to, diag = T)
Mk <- to * subset
colnames(Mk) <- paste0("F(", 1:ncol(Mk) - 1, ")")
# Transform Mk to better suit the table:
Mk_print <- apply(Mk, 2, as.character)
# Create the Tex Table > tabular
Mk.tab <- xtable(Mk_print[1:10, 1:10], digits = 0, format = "latex")
align(Mk.tab) <- rep("r", 11)
```

Mk.tab

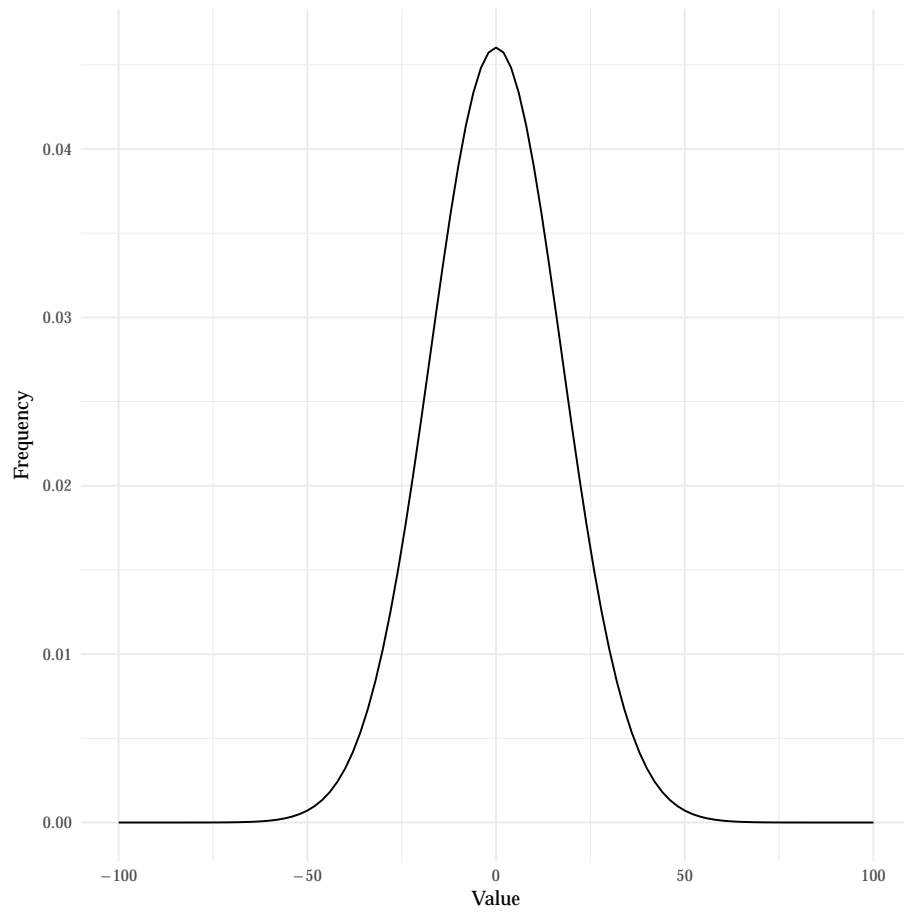
	F(0)	F(1)	F(2)	F(3)	F(4)	F(5)	F(6)	F(7)	F(8)	F(9)
1	0	1	2	3	4	5	6	7	8	9
2	0	-1	0	1	2	3	4	5	6	7
3	0	0	-2	-1	0	1	2	3	4	5
4	0	0	0	-3	-2	-1	0	1	2	3
5	0	0	0	0	-4	-3	-2	-1	0	1
6	0	0	0	0	0	-5	-4	-3	-2	-1
7	0	0	0	0	0	0	-6	-5	-4	-3
8	0	0	0	0	0	0	0	-7	-6	-5
9	0	0	0	0	0	0	0	0	-8	-7
10	0	0	0	0	0	0	0	0	0	-9

```
# compute the probability measure to apply on the Random Variable Mk
fi <- outer(dim_x,
            dim_y,
            FUN = function(i, j){choose((j-1), (j-i)) * p^(j-1)})

range <- 1:ncol(Mk)
lastToss <- ncol(Mk)
# Using ggplot
# data.frame which map distribution and random variable:
distributionSymRanWal <- data.frame(
  Value = Mk[range, lastToss],
  Frequency = fi[range, lastToss]
)

# For the sake of visibility the limit of X axis has been set to [-100, 100]
ggplot(data = distributionSymRanWal, aes(Value, Frequency)) +
  geom_line() +
  scale_x_continuous(limits = c(-100, 100)) +
  theme_minimal() +
  theme(text = element_text(family="CM Roman"),
        axis.title = element_text(face = "plain"))

## Warning: Removed 200 rows containing missing values (geom_path).
```



2 Martingale property

$$\mathbb{E}[X] = \sum_{i=1}^N p_i \times x_i \quad (3)$$

To show the martingale property we have to show that the expectation of the symmetric random walk $\mathbb{E}[X|F(0)] = M_0 = 0$

```
EM200 = sum(Mk[1:200, 200] * fi[1:200, 200]) # equal zero.
```