

# R\_intro\_3



**Karel Fišer, 2017**

# variables & functions

```
> x <- 2
```

```
> y <- c(2,4,5)
```

```
> animals <- "cat"
```

```
> rnorm(7)
```

```
[1] 0.9594941 -0.1102855 -0.5110095 -0.9111954
```

```
[5] -0.8371717 2.4158352 0.1340882
```

```
> random_x <- rnorm(7)
```

```
> rnorm(y[2])
```

# function call

```
> rnorm(7, mean = 10)
```

```
[1]  8.792934 10.277429 11.084441  7.654302
```

```
[5] 10.429125 10.506056  9.425260
```

```
> rnorm(7, mean = 0, sd = 1) # defaults
```

```
> rnorm(7, mean=10, sd=2)
```

```
> rnorm(7, sd=2, mean=10) # with names any order
```

```
> rnorm(3, 10, 2) # w/o names exact order
```

# help(rnorm)

Normal {stats}

R Documentation

## The Normal Distribution

### Description

Density, distribution function, quantile function and random generation for the normal distribution with mean equal to `mean` and standard deviation equal to `sd`.

### Usage

```
dnorm(x, mean = 0, sd = 1, log = FALSE)
pnorm(q, mean = 0, sd = 1, lower.tail = TRUE, log.p = FALSE)
qnorm(p, mean = 0, sd = 1, lower.tail = TRUE, log.p = FALSE)
rnorm(n, mean = 0, sd = 1)
```

### Arguments

`x`, `q` vector of quantiles.

`p` vector of probabilities.

`n` number of observations. If `length(n) > 1`, the length is taken to be the number required.

`mean` vector of means.

`sd` vector of standard deviations.

`log`, `log.p` logical; if TRUE, probabilities `p` are given as  $\log(p)$ .

`lower.tail` logical; if TRUE (default), probabilities are  $P[X \leq x]$  otherwise,  $P[X > x]$ .

## Details

If `mean` or `sd` are not specified they assume the default values of 0 and 1, respectively.

The normal distribution has density

$$f(x) = 1/(\sqrt{2\pi}\sigma) e^{-(x-\mu)^2/(2\sigma^2)}$$

where  $\mu$  is the mean of the distribution and  $\sigma$  the standard deviation.

`qnorm` is based on Wichura's algorithm AS 241 which provides precise results up to about 16 digits.

## Value

`dnorm` gives the density, `pnorm` gives the distribution function, `qnorm` gives the quantile function, and `rnorm` generates random deviates.

The length of the result is determined by `n` for `rnorm`, and is the maximum of the lengths of the numerical parameters for the other functions.

The numerical parameters other than `n` are recycled to the length of the result. Only the first elements of the logical parameters are used.

## Source

For `pnorm`, based on

Cody, W. D. (1993) Algorithm 715: SPECFUN – A portable FORTRAN package of special function routines and test drivers. *ACM Transactions on Mathematical Software* **19**, 22–32.

For `qnorm`, the code is a C translation of

Wichura, M. J. (1988) Algorithm AS 241: The percentage points of the normal distribution. *Applied Statistics*, **37**, 477–484.

For `rnorm`, see [RNG](#) for how to select the algorithm and for references to the supplied methods.

## References

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988) *The New S Language*. Wadsworth & Brooks/Cole.

Johnson, N. L., Kotz, S. and Balakrishnan, N. (1995) *Continuous Univariate Distributions*, volume 1, chapter 13. Wiley, New York.

# help(rnorm)

## See Also

[Distributions](#) for other standard distributions, including [dlnorm](#) for the Lognormal distribution.

# help(rnorm)

## Examples

```
require(graphics)

dnorm(0) == 1/sqrt(2*pi)
dnorm(1) == exp(-1/2)/sqrt(2*pi)
dnorm(1) == 1/sqrt(2*pi*exp(1))

## Using "log = TRUE" for an extended range :
par(mfrow = c(2,1))
plot(function(x) dnorm(x, log = TRUE), -60, 50,
      main = "log { Normal density }")
curve(log(dnorm(x)), add = TRUE, col = "red", lwd = 2)
mtext("dnorm(x, log=TRUE)", adj = 0)
mtext("log(dnorm(x))", col = "red", adj = 1)

plot(function(x) pnorm(x, log.p = TRUE), -50, 10,
      main = "log { Normal Cumulative }")
curve(log(pnorm(x)), add = TRUE, col = "red", lwd = 2)
mtext("pnorm(x, log=TRUE)", adj = 0)
mtext("log(pnorm(x))", col = "red", adj = 1)

## if you want the so-called 'error function'
erf <- function(x) 2 * pnorm(x * sqrt(2)) - 1
## (see Abramowitz and Stegun 29.2.29)
## and the so-called 'complementary error function'
erfc <- function(x) 2 * pnorm(x * sqrt(2), lower = FALSE)
## and the inverses
erfinv <- function(x) qnorm((1 + x)/2)/sqrt(2)
erfcinv <- function(x) qnorm(x/2, lower = FALSE)/sqrt(2)
```

# Rectangular data – data.frame

```
> d <- data.frame(animal=c("cat", "spider"),  
  legs=c(4, 8))  
  
> d[, 2] # d$legs  
  
> d[2, 2] <- 7 # one leg off  
  
  
  
> data(iris) # dataset in R  
  
> write.table(iris, "iris.txt")  
  
> data <- read.table("iris.txt")
```

# I/O

```
> data(iris) # dataset in R
```

```
> write.table(iris, "iris.txt") # where is it?
```

```
> dir()
```

```
> getwd()
```

```
> setwd() # RStudio: Session/Set Working  
Directory/Choose Directory
```

```
> data <- read.table("iris.txt") # use Tab for  
completion
```



# iris.txt

```
iris.txt x
"Sepal.Length" "Sepal.Width" "Petal.Length" "Petal.Width" "Species"
"1" 5.1 3.5 1.4 0.2 "setosa"
"2" 4.9 3 1.4 0.2 "setosa"
"3" 4.7 3.2 1.3 0.2 "setosa"
"4" 4.6 3.1 1.5 0.2 "setosa"
"5" 5 3.6 1.4 0.2 "setosa"
"6" 5.4 3.9 1.7 0.4 "setosa"
"7" 4.6 3.4 1.4 0.3 "setosa"
"8" 5 3.4 1.5 0.2 "setosa"
"9" 4 4 2 9 1 4 0 2 "setosa"
```

	A	B	C	D	E	F	G	H
1	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species			
2	1	5.1	3.5	1.4	0.2	setosa		
3	2	4.9	3	1.4	0.2	setosa		
4	3	4.7	3.2	1.3	0.2	setosa		
5	4	4.6	3.1	1.5	0.2	setosa		
6	5	5	3.6	1.4	0.2	setosa		
7	6	5.4	3.9	1.7	0.4	setosa		
8	7	4.6	3.4	1.4	0.3	setosa		
9	8	5	3.4	1.5	0.2	setosa		
10	9	4 4	2 9	1 4	0 2	setosa		

# iris.txt, iris.csv, iris.xls

empty cell

	A	B	C	D	E	F	G
1		Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	
2	1	5.1	3.5	1.4	0.2	setosa	
3	2	4.9	3	1.4	0.2	setosa	
4	3	4.7	3.2	1.3	0.2	setosa	
5	4	4.6	3.1	1.5	0.2	setosa	
6	5	5	3.6	1.4	0.2	setosa	
7	6	5.4	3.9	1.7	0.4	setosa	
8	7	4.6	3.4	1.4	0.3	setosa	
9	8	5	3.4	1.5	0.2	setosa	
10	9	4.4	3.0	1.4	0.2	setosa	

```
> read.table("iris.txt")
```

```
> read.table("iris.csv", sep=";", header=TRUE,  
row.names=1)
```

```
> library(xlsx)
```

```
> read.xlsx("iris.xls")
```

# read.table()

```
header = FALSE,  
sep = ",",  
quote = "\"'",  
dec = ".",  
row.names, col.names,  
...  
check.names = TRUE,  
fill = !blank.lines.skip,  
...  
stringsAsFactors = default.stringsAsFactors()  
...
```

# read.table - SAM

```
> sam <- read.table("LB-7K.final.sam", fill=TRUE,  
  nrows=1000000)
```

```
> sam[255000, 10]
```

```
[1]
```

```
"CCGGGGACTCCATCAAGGCCATCGCATCCATTGCCGACAANGANNNNNNN  
NNCANNNTCATGGGGATGCAGGAGAAGAAGCTGGGCTCGCTCCCCTACCAC  
"
```

```
> sam[255000, 11]
```

```
[1] "8@<D;D:@FFHBH<E?<@93:??1CGDG>FFGHG"
```

# read.table → t.test

```
> data <- read.table("iris.txt")  
> t.test(data[,1], data[,3])
```

Welch Two Sample t-test

data: data[, 1] and data[, 3]

t = 13.0984, df = 211.543, p-value < 2.2e-16

alternative hypothesis: true difference in means is not equal to 0

95 percent confidence interval:

1.771500 2.399166

sample estimates:

mean of x mean of y

5.843333 3.758000

# Your turn

create data outside R and load it in

# packages

```
> install.packages("ggplot2") # CRAN  
> # in RStudio Packages/Install Packages  
  
> source("http://bioconductor.org/biocLite.R")  
> biocLite("flowCore") # Bioconductor  
  
> library(devtools)  
> install_github("username/packageName") # github  
  
> library(ggplot2)
```

# packages - info

```
> library(help=ggplot2) # description  
> library(ggplot2); ls("package:ggplot2")  
> data(package="ggplot2")
```

where packages live:

```
> .libPaths()  
> find.package("ggplot2", lib.loc=NULL, quiet =  
TRUE)
```



# packages – where to start

- Google
- PubMed, StackOverflow
  
- CRAN and Bioconductor views
  
- vignettes
- (manuals), help
- explore the code (functions)

## CRAN Task Views

<a href="#">Bayesian</a>	Bayesian Inference
<a href="#">ChemPhys</a>	Chemometrics and Computational Physics
<a href="#">ClinicalTrials</a>	Clinical Trial Design, Monitoring, and Analysis
<a href="#">Cluster</a>	Cluster Analysis & Finite Mixture Models
<a href="#">DifferentialEquations</a>	Differential Equations
<a href="#">Distributions</a>	Probability Distributions
<a href="#">Econometrics</a>	Computational Econometrics
<a href="#">Environmetrics</a>	Analysis of Ecological and Environmental Data
<a href="#">ExperimentalDesign</a>	Design of Experiments (DoE) & Analysis of Experimental Data
<a href="#">Finance</a>	Empirical Finance
<a href="#">Genetics</a>	Statistical Genetics
<a href="#">Graphics</a>	Graphic Displays & Dynamic Graphics & Graphic Devices & Visualization
<a href="#">HighPerformanceComputing</a>	High-Performance and Parallel Computing with R
<a href="#">MachineLearning</a>	Machine Learning & Statistical Learning
<a href="#">MedicalImaging</a>	Medical Image Analysis
<a href="#">MetaAnalysis</a>	Meta-Analysis
<a href="#">Multivariate</a>	Multivariate Statistics
<a href="#">NaturalLanguageProcessing</a>	Natural Language Processing
<a href="#">NumericalMathematics</a>	Numerical Mathematics
<a href="#">OfficialStatistics</a>	Official Statistics & Survey Methodology
<a href="#">Optimization</a>	Optimization and Mathematical Programming
<a href="#">Pharmacokinetics</a>	Analysis of Pharmacokinetic Data
<a href="#">Phylogenetics</a>	Phylogenetics, Especially Comparative Methods
<a href="#">Psychometrics</a>	Psychometric Models and Methods
<a href="#">ReproducibleResearch</a>	Reproducible Research
<a href="#">Robust</a>	Robust Statistical Methods
<a href="#">SocialSciences</a>	Statistics for the Social Sciences
<a href="#">Spatial</a>	Analysis of Spatial Data
<a href="#">SpatioTemporal</a>	Handling and Analyzing Spatio-Temporal Data
<a href="#">Survival</a>	<a href="#">Survival Analysis</a>
<a href="#">TimeSeries</a>	Time Series Analysis
<a href="#">WebTechnologies</a>	Web Technologies and Services
<a href="#">gR</a>	gRaphical Models in R

# vignette

## Using Time Dependent Covariates and Time Dependent Coefficients in the Cox Model

Terry Therneau      Cindy Crowson  
Mayo Clinic

February 26, 2013

### 1 Introduction

One of the strengths of the Cox model is its ability to encompass covariates that change over time, due to the theoretical foundation in martingales. A *martingale* (original definition) is a betting strategy in games of chance. One of the simplest and best known is doubling the bet each time you lose. For instance consider the following game of roulette:

Bet	Outcome	Win	Running total
R \$1	Red	2	1
R \$1	Black	0	0
R \$2	Black	0	-2
B \$4	Red	0	-6
R \$8	Black	0	-14
B \$16	Black	32	2
B \$1	Red	0	1
B \$2	Black	4	3
:	:	:	:

At the end of each cycle of bets the player is another \$1 ahead. The problem is that a modest sequence of losses will exhaust their stake.

The rule for time dependent covariates in a Cox model is simple and essentially the same as that for gambling: you cannot look into the future. A covariate may change in any way based on past data or outcomes, but it may not reach "forward" in time. One of the more well known examples of this error is analysis by response: at the end of a trial a survival curve is made comparing those who had an early response to treatment (shrinkage of tumor, lowering of cholesterol, or whatever), and it discovered that response predicts survival. The problem arises because subjects are classified as responders or non-responders from the beginning of the study, i.e., they are placed into group A or B before the response has occurred. As a consequence, any

# save your work

## objects:

```
> save(x, y, file="xy.rda")  
> # saveRDS(x, "x.rds")  
> save.image() # "work.RData"  
> load("xy.rda")
```

## data:

```
> write.table(); read.table()
```

## code:

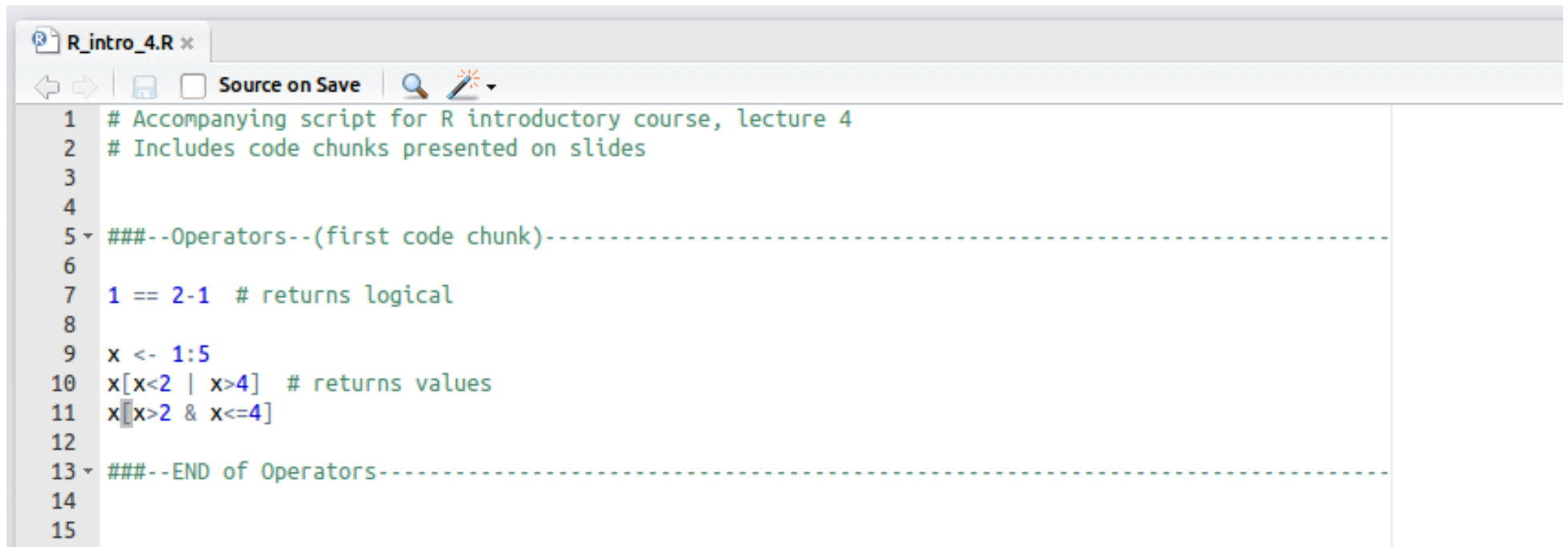
```
> # use editor to save scripts  
> source("iris_plots.R")
```

```
$ R CMD BATCH [options] iris_plots.R [outfile]
```

# scripts

Scripts are plain text files storing your code.

Store them with ending .R (e.g. r\_intro.R)



```
R_intro_4.R x
Source on Save
1 # Accompanying script for R introductory course, lecture 4
2 # Includes code chunks presented on slides
3
4
5 ▾ ###--Operators--(first code chunk)-----
6
7 1 == 2-1 # returns logical
8
9 x <- 1:5
10 x[x<2 | x>4] # returns values
11 x[x>2 & x<=4]
12
13 ▾ ###--END of Operators-----
14
15
```

# scripts

Scripts are plain text files storing your code.

- repeat your work any time → reproducible research
- repeat your work with any data → reports
- monitor changes → code versions
  
- have a style (e.g. google-r-style.html)
- make comments! # make the code understandable to others (e.g. your future self)

# make comments

```
> # comment why not what (with exceptions)
```

```
> # whole line or
```

```
> x <- 2 # after code
```

```
>
```

```
> # separate code into chunks:
```

```
> # -----
```

```
> # =====
```

```
> # ---- Title -----
```

```
>
```

```
> # TODO (who)
```

# Rstudio - shortcuts

Alt + Shift + K

- Ctrl + Enter
- Ctrl + 1, Ctrl + 2
- Ctrl + → , Ctrl + Shift + →
- Ctrl + Shift + c
- Alt + -
- Esc
- Tab
- up-arrow, or Ctrl + R

<https://www.rstudio.com/resources/cheatsheets/>



# RStudio

The screenshot displays the RStudio environment. The top menu bar includes File, Edit, Code, View, Plots, Session, Build, Debug, Profile, Tools, and Help. Below the menu is a toolbar with icons for file operations and a search bar. The main editor window shows a script named 'test\_script.R' with the following code:

```
1 # Write your code into scripts
2 x <- c(2, 3, 10:15)
3 |
```

The console pane at the bottom left shows the R version (3.3.1) and the execution of the code from the script, resulting in the output:

```
> x <- c(2, 3, 10:15)
> |
```

The environment pane on the right shows the 'Global Environment' with a search bar and a 'Values' section displaying the variable 'x' as a numeric vector of length 8: 'num [1:8] 2 3 10 11 12 13 14 15'. The bottom right pane shows the 'Viewer' tab with the R documentation for 'read.table', including sections for 'Description' and 'Usage'.

# Your turn

write a script

<http://www.bioinformin.net>

