

Essential R Commands for Statistical Work

1. Basic R Commands

Purpose	Command	Example
Check working directory	<code>getwd()</code>	
Set working directory	<code>setwd("path/to/folder")</code>	
List objects in memory	<code>ls()</code>	
Remove object	<code>rm(x)</code>	
Help for a function	<code>?mean</code> or <code>help(mean)</code>	
Install a package	<code>install.packages("dplyr")</code>	
Load a package	<code>library(dplyr)</code>	

2. Syntax and Working with Functions

Purpose	Command / Example	Description
Assignment	<code>x <- 10</code>	Create a variable
Comment	<code># This is a comment</code>	
Vector	<code>x <- c(1, 2, 3, 4)</code>	Basic data structure
Sequence	<code>1:10</code> or <code>seq(1, 10, by=2)</code>	Generate sequences
Repeat values	<code>rep(5, times=3)</code>	Result: 5 5 5
Vector of 1000 of 1	<code>x <- rep(1, times = 1000)</code>	
Check object type	<code>class(x)</code> or <code>typeof(x)</code>	
Length of object	<code>length(x)</code>	
Conditional expression	<code>if (x > 5) {print("yes")} else {print("no")}</code>	
Loop	<code>for (i in 1:5) print(i)</code>	
Apply function to vector	<code>sapply(x, sqrt)</code>	

Creating custom functions:

```
my_function <- function(a, b = 2) {  
  result <- a * b  
  return(result)  
}
```

```
my_function(3) # 6
```

3. Basic Statistical Functions

Purpose	Command	Example
Mean	<code>mean(x)</code>	
Median	<code>median(x)</code>	
Standard deviation	<code>sd(x)</code>	
Variance	<code>var(x)</code>	
Quantiles	<code>quantile(x)</code>	
Correlation	<code>cor(x, y)</code>	
Covariance	<code>cov(x, y)</code>	

4. Basic Visualization

Purpose	Command	Example
Histogram	<code>hist(x)</code>	
Scatter plot	<code>plot(x, y)</code>	
Boxplot	<code>boxplot(x)</code>	
Bar plot	<code>barplot(table(x))</code>	
Save plot to PDF	<code>pdf("plot.pdf"); plot(x, y); dev.off()</code>	

Modeling of random variables.

Simulation of Random Variables in R

1. Direct Simulation (Built-in Distributions)

R provides built-in random number generators for most common probability distributions. They all follow the same naming convention:

$$r < \textit{distribution} > (n, \text{parameters})$$

where **n** is the number of random values to generate.

Distribution	Function	Example
Uniform	<code>runif()</code>	<code>runif(1000, min = 0, max = 1)</code>
Normal	<code>rnorm()</code>	<code>rnorm(1000, mean = 0, sd = 1)</code>
Bernoulli / Binomial	<code>rbinom()</code>	<code>rbinom(1000, size = 1, prob = 0.5)</code>
Poisson	<code>rpois()</code>	<code>rpois(1000, lambda = 4)</code>
Exponential	<code>rexp()</code>	<code>rexp(1000, rate = 1/2)</code>
Chi-square	<code>rchisq()</code>	<code>rchisq(1000, df = 5)</code>
Student's t	<code>rt()</code>	<code>rt(1000, df = 10)</code>
F distribution	<code>rf()</code>	<code>rf(1000, df1 = 5, df2 = 10)</code>
Beta	<code>rbeta()</code>	<code>rbeta(1000, 2, 5)</code>
Gamma	<code>rgamma()</code>	<code>rgamma(1000, shape = 2, rate = 1)</code>

2. Inverse Transform Method

This method is used when the cumulative distribution function (CDF) $F(x)$ is known, but there is no direct generator for the random variable.

Idea:

1. Generate $U \sim \text{Uniform}(0, 1)$.
2. Compute $X = F^{-1}(U)$, where F^{-1} is the inverse CDF.

Example: Simulating an Exponential Distribution manually

```
# Inverse Transform Sampling for an Exponential RV with rate = 2
u <- runif(1000)
x <- -log(1 - u) / 2
hist(x, main = "Simulated Exponential Distribution", col = "lightblue")
```

The generated variable **x** now follows an exponential distribution with rate parameter $\lambda = 2$.

3. Working with Vectors in R

Vectors are the fundamental data structure in R. They can store numeric, character, or logical values and allow vectorized operations.

Creating vectors:

```
x <- c(1, 2, 3, 4, 5)      # combine values into a vector
y <- rep(1, 1000)          # 1000 ones
z <- seq(0, 10, by = 2)    # sequence: 0, 2, 4, 6, 8, 10
```

Basic operations:

```
x + 2      # adds 2 to each element
x * y      # element-wise multiplication
x[3]       # third element
x[1:3]     # first three elements
```

The sum() function: `sum()` adds up all the elements of a numeric vector and returns their total. It is a fully vectorized operation, meaning it automatically processes each element.

```
x <- c(1, 2, 3, 4)
sum(x)      # 10
sum(x > 2)  # counts how many elements are greater than 2
sum(is.na(x)) # counts missing values (NA)
```

Note: Because R is vectorized, expressions like `mean(x)`, `var(x)`, and `sd(x)` also work directly on vectors without loops.