

# UNIT-I

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**Basics of R:** Introduction, R-Environment Setup, Programming with R, Basic Data Types. **Vectors:** Creating and Naming Vectors, Vector Arithmetic, Vector Subsetting. **Matrices:** Creating and Naming Matrices, Matrix Subsetting. **Arrays, Class.**

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## 1. What is R?

R is an **open-source programming language and software environment** used for:

- Statistical computing
- Data analysis
- Data visualization
- Scientific research

R was developed by **Ross Ihaka and Robert Gentleman** at the **University of Auckland, New Zealand**.

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## 2. Why R is Used?

R is widely used because it provides:

- Powerful statistical functions
- Advanced graphical capabilities
- Thousands of ready-to-use packages
- Easy data handling and manipulation

It is especially popular among **data scientists, statisticians, researchers, and academicians**.

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## 3. Features of R

- **Free and Open Source**
- **Platform Independent** (Windows, Linux, macOS)
- Supports **Object-Oriented Programming**
- Large package ecosystem (CRAN)
- Excellent **data visualization**
- Strong community support
- Interpreted language (no compilation needed)

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#### 4. Applications of R

R is used in many domains such as:

- **Data Science & Analytics**
  - **Machine Learning**
  - **Bioinformatics**
  - **Finance & Banking**
  - **Healthcare Analytics**
  - **Academic Research**
  - **Artificial Intelligence**
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#### 5. Advantages of R

- Easy to learn and use
  - High-quality graphics and plots
  - Large number of statistical techniques
  - Easy integration with other languages (Python, C, Java)
  - Frequent updates and improvements
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#### 6. Limitations of R

- Slower execution for very large datasets
  - Memory-intensive (stores data in RAM)
  - Not ideal for large-scale production systems
  - Steep learning curve for beginners in programming
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#### 7. R as a Programming Language

- R is an **interpreted language**
- Commands are executed line by line
- Case-sensitive (x and X are different)
- Supports **procedural, functional, and OOP styles**

Example:

```
x <- 10
```

```
y <- 20
```

```
x + y
```

---

## 8. R Packages

- Packages are collections of functions
- Available through **CRAN**
- Examples: ggplot2, dplyr, tidyverse

Installing a package:

```
install.packages("ggplot2")
```

```
library(ggplot2)
```

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## 9. R Environment Components

- **Console** – Executes commands
  - **Script Editor** – Write programs
  - **Workspace** – Stores objects
  - **Packages** – Additional functionalities
- 

## 10. Comparison: R vs Other Languages

Feature	R	Python
Primary use	Statistics	General purpose
Visualization	Excellent	Good
Learning curve	Moderate	Easy
Packages	CRAN	PyPI

## Vectors

### 1. What is a Vector in R?

A **vector** is the **most basic data structure in R**.

It is a **one-dimensional collection of elements** of the **same data type**.

✓ All elements in a vector must be **homogeneous** (same type).

## Feature

## R

## Python

### 2. Creating Vectors in R

#### 2.1 Using c() Function (Combine)

```
v1 <- c(10, 20, 30, 40)
```

```
v2 <- c("R", "Python", "Java")
```

```
v3 <- c(TRUE, FALSE, TRUE)
```

Check type:

```
class(v1)
```

#### 2.2 Using : Operator (Sequence)

```
v <- 1:10
```

#### 2.3 Using seq() Function

```
v <- seq(from = 1, to = 10, by = 2)
```

#### 2.4 Using rep() Function

```
v <- rep(5, times = 4)
```

```
v2 <- rep(c(1, 2), times = 3)
```

### 3. Naming Vectors

#### 3.1 Naming During Creation

```
marks <- c(Maths = 85, Physics = 78, Chemistry = 90)
```

#### 3.2 Naming After Creation

```
marks <- c(85, 78, 90)
```

```
names(marks) <- c("Maths", "Physics", "Chemistry")
```

Access named elements:

```
marks["Maths"]
```

Feature	R	Python
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### 3.3 Removing Names

```
names(marks) <- NULL
```

---

## 4. Vector Arithmetic

R supports **element-wise operations** on vectors.

### 4.1 Addition, Subtraction, Multiplication, Division

```
a <- c(10, 20, 30)
```

```
b <- c(1, 2, 3)
```

```
a + b
```

```
a - b
```

```
a * b
```

```
a / b
```

---

### 4.2 Scalar Operations

```
a <- c(10, 20, 30)
```

```
a + 5
```

```
a * 2
```

---

### 4.3 Vector Recycling Rule

If vectors are of different lengths, R **recycles** elements.

```
x <- c(1, 2, 3, 4)
```

```
y <- c(10, 20)
```

```
x + y
```

---

### 4.4 Logical Operations on Vectors

```
x <- c(5, 10, 15)
```

```
x > 8
```

## 5. Vector Subsetting (Accessing Elements)

Subsetting means **selecting specific elements** from a vector.

### 5.1 By Index (Position)

```
v <- c(10, 20, 30, 40, 50)
```

```
v[1]    # First element
```

```
v[3]    # Third element
```

```
v[c(2,4)] # Multiple elements
```

### 5.2 Negative Indexing (Excluding Elements)

```
v[-1]    # Exclude first element
```

```
v[-c(2,3)] # Exclude 2nd and 3rd elements
```

### 5.3 Logical Subsetting

```
v <- c(10, 20, 30, 40)
```

```
v[v > 25]
```

### 5.4 Subsetting Using Names

```
marks <- c(Maths = 85, Physics = 78, Chemistry = 90)
```

```
marks["Physics"]
```

### 5.5 Using which() Function

```
v <- c(5, 12, 18, 7)
```

```
v[which(v > 10)]
```

## 6. Useful Vector Functions

### Function Purpose

length() Number of elements

sum() Sum of elements

mean() Average

min() Minimum

max() Maximum

sort() Sorting

unique() Unique values

Example:

```
v <- c(10, 20, 20, 30)
```

```
unique(v)
```

## 7. Type Coercion in Vectors

```
v <- c(10, "R", TRUE)
```

```
class(v)
```

✓ All elements are converted to **character**

# Matrices

## 1. What is a Matrix in R?

A **matrix** is a **two-dimensional data structure** in R that stores elements in **rows and columns**.

✓ All elements in a matrix must be of the **same data type** (homogeneous).

## 2. Creating Matrices in R

### 2.1 Using matrix() Function

**Syntax:**

```
matrix(data, nrow, ncol, byrow = FALSE)
```

**Example:**

```
m1 <- matrix(c(1, 2, 3, 4, 5, 6), nrow = 2, ncol = 3)
```

```
m1
```

(Default filling is **column-wise**)

---

**2.2 Creating Matrix by Row**

```
m2 <- matrix(c(1, 2, 3, 4, 5, 6), nrow = 2, byrow = TRUE)
```

---

**2.3 Creating Matrix Using Vectors**

```
v1 <- c(1, 2, 3)
```

```
v2 <- c(4, 5, 6)
```

```
m <- cbind(v1, v2) # Column bind
```

```
m
```

```
m <- rbind(v1, v2) # Row bind
```

```
m
```

---

**2.4 Creating Special Matrices**

```
matrix(0, nrow = 3, ncol = 3) # Zero matrix
```

```
matrix(1, nrow = 2, ncol = 2) # Ones matrix
```

---

**3. Naming Rows and Columns of a Matrix****3.1 Naming During Creation**

```
m <- matrix(
```

```
  c(10, 20, 30, 40),
```

```
  nrow = 2,
```

```
  dimnames = list(
```



```
c("Row1", "Row2"),  
c("Col1", "Col2")  
)  
)
```

---

### 3.2 Naming After Creation

```
rownames(m) <- c("R1", "R2")  
colnames(m) <- c("C1", "C2")
```

---

### 3.3 Viewing Names

```
rownames(m)  
colnames(m)
```

---

### 3.4 Removing Names

```
rownames(m) <- NULL  
colnames(m) <- NULL
```

---

## 4. Accessing Matrix Elements (Matrix Subsetting)

Matrix subsetting uses:

```
matrix[row, column]
```

---

### 4.1 Access Single Element

```
m <- matrix(1:9, nrow = 3)
```

```
m[2, 3] # 2nd row, 3rd column
```

---

### 4.2 Access Entire Row or Column

```
m[1, ] # First row
```

```
m[, 2] # Second column
```

---

#### 4.3 Access Multiple Rows and Columns

```
m[c(1,3), c(2,3)]
```

---

#### 4.4 Excluding Rows or Columns (Negative Indexing)

```
m[-1, ] # Remove first row
```

```
m[, -2] # Remove second column
```

---

#### 4.5 Subsetting Using Names

```
rownames(m) <- c("A", "B", "C")
```

```
colnames(m) <- c("X", "Y", "Z")
```

```
m["B", "Y"]
```

---

#### 4.6 Logical Subsetting

```
m[m > 5]
```

---

### 5. Matrix Operations (Basic)

#### 5.1 Element-wise Operations

```
m1 <- matrix(1:4, nrow = 2)
```

```
m2 <- matrix(5:8, nrow = 2)
```

```
m1 + m2
```

```
m1 * m2
```

---

#### 5.2 Matrix Multiplication

```
m1 %*% m2
```

---

#### 5.3 Transpose of Matrix

t(m1)

---

## 6. Useful Matrix Functions

Function	Purpose
----------	---------

dim()	Dimensions
-------	------------

nrow()	Number of rows
--------	----------------

ncol()	Number of columns
--------	-------------------

rowSums()	Sum of rows
-----------	-------------

colSums()	Sum of columns
-----------	----------------

apply()	Apply function
---------	----------------

Example:

```
rowSums(m)
```

---

## 7. Type Coercion in Matrices

```
m <- matrix(c(1, 2, "R", 4), nrow = 2)
```

```
class(m)
```

✓ All values become **character**

# Arrays

## 1. What is an Array in R?

An **array** is a **multidimensional data structure** in R that can store data in **more than two dimensions**.

- A **vector** → 1D
- A **matrix** → 2D
- An **array** → 2D or more (3D, 4D, etc.)

✓ All elements in an array must be of the **same data type** (homogeneous).

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## 2. Difference Between Matrix and Array

Feature	Matrix	Array
Dimensions	2D	2D or more
Rows & Columns	Yes	Yes
Layers	No	Yes
Data type	Homogeneous	Homogeneous

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### 3. Creating Arrays in R

#### 3.1 Using array() Function

**Syntax:**

```
array(data, dim, dimnames = NULL)
```

---

#### 3.2 Creating a 2D Array

```
a <- array(1:6, dim = c(2, 3))
```

```
a
```

✓ Filled **column-wise** by default.

---

#### 3.3 Creating a 3D Array

```
a3 <- array(1:12, dim = c(2, 3, 2))
```

```
a3
```

- 2 rows
  - 3 columns
  - 2 matrices (layers)
- 

#### 3.4 Creating Array Using Vectors

```
v <- c(10, 20, 30, 40)
```

```
arr <- array(v, dim = c(2, 2, 1))
```

```
arr
```

---

### 4. Naming Dimensions of an Array

#### 4.1 Naming While Creation

```
arr <- array(
  1:8,
  dim = c(2, 2, 2),
  dimnames = list(
    c("R1", "R2"),
    c("C1", "C2"),
    c("Layer1", "Layer2")
  )
)
arr
```

---

## 4.2 Naming After Creation

```
dimnames(arr) <- list(
  c("Row1", "Row2"),
  c("Col1", "Col2"),
  c("M1", "M2")
)
```

---

## 5. Accessing Elements in an Array (Subsetting)

### General Format:

```
array[row, column, layer]
```

---

### 5.1 Access a Single Element

```
arr[1, 2, 1]
```

---

### 5.2 Access Entire Row, Column, or Layer

```
arr[1, , ] # First row
```

```
arr[ , 2, ] # Second column
```

```
arr[ , , 1] # First layer
```

---

### 5.3 Access Multiple Elements

```
arr[c(1,2), c(1,2), 1]
```

---

#### 5.4 Subsetting Using Names

```
arr["Row1", "Col2", "M1"]
```

---

#### 5.5 Logical Subsetting

```
arr[arr > 5]
```

---

### 6. Array Operations

#### 6.1 Element-wise Arithmetic

```
a1 <- array(1:8, dim = c(2,2,2))
```

```
a2 <- array(9:16, dim = c(2,2,2))
```

```
a1 + a2
```

```
a1 * a2
```

---

#### 6.2 Apply Functions on Arrays

```
apply(arr, c(1), sum) # Row-wise
```

```
apply(arr, c(2), mean) # Column-wise
```

```
apply(arr, c(3), sum) # Layer-wise
```

---

### 7. Useful Array Functions

Function	Purpose
dim()	Dimensions
length()	Total elements
dimnames()	Dimension names
apply()	Apply function
str()	Structure

Example:

```
dim(arr)
```

---

## 8. Type Coercion in Arrays

```
arr <- array(c(1, 2, "R", 4), dim = c(2,2))
```

```
class(arr)
```

✓ All values become **character**

# Class

## 1. What is a Class in R?

In R, a **class** defines the **type and structure of an object**.

It tells R **how data should be treated** and **which methods (functions) apply to it**.

◆ Every object in R belongs to at least **one class**.

Example:

```
x <- 10
```

```
class(x)
```

Output:

```
[1] "numeric"
```

---

## 2. Importance of Class in R

- Determines how functions behave on an object
- Enables **Object-Oriented Programming (OOP)**
- Helps in **data organization**
- Improves code reusability and readability

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## 3. Built-in Classes in R

Data	Class
10	numeric
10L	integer
"R"	character

<b>Data</b>	<b>Class</b>
TRUE	logical
2+3i	complex
c(1,2,3)	numeric
list()	list
matrix()	matrix
data.frame()	data.frame
factor()	factor

---

#### 4. Checking and Assigning Class

##### 4.1 Check Class

```
x <- c(1, 2, 3)
class(x)
```

---

##### 4.2 Assign a Class

```
class(x) <- "MyClass"
class(x)
```

---

##### 4.3 Multiple Classes

```
class(x) <- c("MyClass", "numeric")
```

---

#### 5. Object-Oriented Programming Systems in R

R supports **four OOP systems**:

##### OOP System Description

<b>S3</b>	Simple, informal, most commonly used
<b>S4</b>	Formal, strict definitions
<b>R6</b>	Reference-based, modern OOP
<b>RC</b>	Older reference classes

---



## 6. S3 Class (Most Important for Exams)

### 6.1 Creating an S3 Class

#### Step 1: Create Object

```
student <- list(  
  name = "Ravi",  
  age = 20,  
  branch = "CSE"  
)
```

#### Step 2: Assign Class

```
class(student) <- "Student"
```

---

### 6.2 Constructor Function

```
Student <- function(name, age, branch) {  
  obj <- list(  
    name = name,  
    age = age,  
    branch = branch  
  )  
  class(obj) <- "Student"  
  return(obj)  
}
```

```
s1 <- Student("Anita", 21, "IT")
```

---

## 7. Methods in S3 Class

A **method** is a function that works for a specific class.

#### Example: print() Method

```
print.Student <- function(x) {  
  cat("Student Details\n")  
  cat("Name:", x$name, "\n")  
  cat("Age:", x$age, "\n")  
}
```

```
cat("Branch:", x$branch, "\n")
}
```

```
print(s1)
```

---

## 8. Generic Functions and Method Dispatch

- `print()` is a **generic function**
- R automatically calls `print.ClassName()`

```
print(s1) # Calls print.Student()
```

---

## 9. S4 Class (Brief Overview)

S4 classes are **formal and strict**.

**Example:**

```
setClass(
  "Employee",
  slots = list(
    id = "numeric",
    name = "character",
    salary = "numeric"
  )
)
```

```
emp <- new("Employee", id = 101, name = "Ravi", salary = 50000)
```

---

## 10. R6 Class (Brief Overview)

R6 uses **reference semantics**.

```
library(R6)
```

```
Person <- R6Class("Person",
  public = list(
    name = NULL,
```

```

initialize = function(name) {
  self$name <- name
},
greet = function() {
  cat("Hello,", self$name)
}
)
)

```

```

p <- Person$new("Ravi")
p$greet()

```

## 11. Class vs Data Type

Feature	Data Type	Class
Meaning	Kind of data	Structure + behavior
Example	numeric	data.frame
Purpose	Storage	OOP

### Difference Between S3 and R6 Classes

Feature	S3 Class	R6 Class
OOP type	Informal	Formal
Definition	No strict definition	Strict class definition
Object behavior	Copy-on-modify	Modified in place
Method calling	print(obj)	obj\$method()
Package required	No	Yes (R6)
Ease of use	Very easy	Moderate