# Matrices

# Chapter 4

# Stats 20 Lec 2

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# Learning Objectives

After studying this chapter, you should be able to:

- Create matrices with matrix(), rbind(), and cbind().
- Understand how R stores and performs operations on matrices.
- Add row and column names to two-dimensional objects.
- Extract and assign values to two-dimensional objects.
- Differentiate between \* and %\*% for matrices.
- Compute the inverse of a square matrix.

### 1 Basic Definitions and Functions

Matrices play an important role in statistics, especially in multivariate analyses, such as multiple regression or estimating the covariance matrix for a collection of random variables (i.e., a random vector). We will not focus on linear algebra or matrix theory in this course, but we will introduce the matrix object in R and some important concepts related to them. Matrices provide intuition and a natural segue into working with data frames, possibly the most commonly used object in R to store rectangular data. Most of the syntax and functions that apply to matrices will also apply to data frames.

### 1.1 The matrix() Function

A matrix is a two-dimensional (rectangular) array of values. In R, every value in a matrix must be of the same type (i.e., all numeric, character, logical). The matrix() function takes in a vector of values (called data) and arranges them into a matrix according to the desired number of rows (nrow) and number of columns (ncol).

```
A <- matrix(1:6,nrow=2,ncol=3)

## [,1] [,2] [,3]

## [1,] 1 3 5

## [2,] 2 4 6
```

Note: By default, the matrix() function reads the input vector and fills in the matrix by column. For example, in matrix A, the first two elements fill in the first column, the next two elements in the second column, and the last two elements fill in the last column. To fill the matrix in by row, set the optional byrow argument to TRUE (the default value is byrow=FALSE, which means R will fill the matrix by column).

```
B <- matrix(1:9,nrow=3,ncol=3,byrow=TRUE)
B</pre>
```

```
## [,1] [,2] [,3]
## [1,] 1 2 3
## [2,] 4 5 6
## [3,] 7 8 9
```

**Note**: The default values are nrow=1 and ncol=1. Leaving both blank will produce a matrix with a single column. If only the nrow or ncol argument is defined, the argument left blank will be automatically computed based on the length of the data vector.

Caution: If the length of the data vector is too short to fill the entire matrix, the values of data will be recycled. The way R behaves when recycling in this context is similar to the vector recycling from Chapter 2.

If the vector is recycled a whole number of times, R will fill the matrix with no warning that it has recycled the data values. If the vector is not recycled completely, the matrix will still be filled, but R will also throw a warning.

```
matrix(1:4,nrow=2,ncol=6) # Complete recycling: no warning thrown
        [,1] [,2] [,3] [,4] [,5] [,6]
## [1,]
                3
                     1
                           3
                                1
           1
## [2,]
           2
                4
                      2
matrix(1:4,nrow=2,ncol=5) # Incomplete recycling: warning thrown
## Warning in matrix(1:4, nrow = 2, ncol = 5): data length [4] is not a sub-
## multiple or multiple of the number of columns [5]
        [,1] [,2] [,3] [,4] [,5]
##
## [1,]
           1
                3
                           3
                      1
## [2,]
           2
                      2
                                2
```

### 1.2 The Dimension of a Matrix

The **dimension** of a matrix (or any two-dimensional object) is the number of rows and the number of columns, often written as nrow  $\times$  ncol, read aloud as "nrow by ncol". For example, matrix A is a  $2 \times 3$  (2 by 3) matrix. A matrix is called **square** if the number of rows equals the number of columns.

The dim() function returns a numeric vector specifying the dimension of the input object.

```
dim(A) # dimension of A

## [1] 2 3
dim(B) # dimension of B

## [1] 3 3
```

Caution: The order matters in the dimension of a matrix. A  $2 \times 3$  matrix has a different dimension from a  $3 \times 2$  matrix. The number of rows is always listed first, and the number of columns is always listed second.

The functions **nrow()** and **ncol()** input a two-dimensional object and output the number of rows or columns, respectively. These will produce the individual entries from the **dim()** function.

```
nrow(A) # The number of rows of A
## [1] 2
ncol(B) # The number of columns of B
## [1] 3
```

Question: How would you extract the number of rows from the output of dim(A)?

#### 1.3 The cbind() and rbind() Functions

Another way to create a matrix is to use the **cbind()** and **rbind()** functions, which combine ("bind") columns and rows (respectively) into a matrix. Input each column or row values as a separate argument, as shown below:

```
# An alternative way to create the matrix A cbind(c(1,2),c(3,4),c(5,6))
```

```
## [,1] [,2] [,3]
## [1,] 1 3 5
## [2,] 2 4 6
# An alternative way to create the matrix B
rbind(c(1,2,3),c(4,5,6),c(7,8,9))
## [,1] [,2] [,3]
## [1,] 1 2 3
```

If the number of columns or rows match in a way that makes sense, i.e., they are **conformable**, then **rbind()** and **cbind()** can also input two-dimensional objects and bind them together. Trying to combine matrices that are not conformable will produce an error.

```
# Bind the rows of A and B together (i.e., stack A on top of B)
rbind(A,B)
##
        [,1] [,2] [,3]
## [1,]
           1
                 3
## [2,]
           2
                      6
                      3
## [3,]
           1
                 2
## [4,]
           4
                 5
                      6
           7
                      9
## [5,]
# This will give an error, because the number of rows are not conformable
```

```
## Error in cbind(A, B): number of rows of matrices must match (see arg 2)
```

Side Note: The rbind() and cbind() functions also use recycling when necessary. This is particularly useful when inserting a column or row of repeated values. For an application to statistics: In linear regression (especially multiple regression), the observed values of the predictor variable(s) are often organized into a matrix, called the design matrix. The design matrix typically has a column of 1's to account for the intercept term in the linear model. An example of appending a column of 1's is given below.

```
cbind(1,B) # Append a column of 1's to matrix B
## [,1] [,2] [,3] [,4]
```

```
## [1,] 1 1 2 3
## [2,] 1 4 5 6
## [3,] 1 7 8 9
```

#### 1.4 Matrices Are Two-Dimensional Vectors

The reason why every value in a matrix must be of the same type is because all matrix objects are actually internally stored as vectors in R. This can be verified by using the mode() function on a matrix.

```
mode(A) # A is stored as a numeric vector
```

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

## [2,]

## [3,]

cbind(A,B)

4

5

6

Matrices are just vectors with an additional **attribute** called dimension (dim). Vectors have no dimension attribute (i.e., vectors are not just one-dimensional matrices).

The attributes() function allows us to see the attributes of an R object.

```
attributes(A)
```

```
## $dim
## [1] 2 3
attributes(1:6)
```

## NULL

We could strip the A matrix of its dim attribute by assigning NULL to its attributes (A) object. The matrix object will revert back to a vector.

```
attributes(A) <- NULL # Remove all of A's attributes
A # A is now a vector</pre>
```

```
## [1] 1 2 3 4 5 6
```

We can also give a vector the dim attribute in a similar way to convert a vector into a matrix.

```
# Assign the dim attribute to A (with 2 rows and 3 columns)
attributes(A) <- list(dim=c(2,3))
A # A is a matrix again</pre>
```

```
## [,1] [,2] [,3]
## [1,] 1 3 5
## [2,] 2 4 6
```

# 2 Naming Rows and Columns of Two-Dimensional Objects

Suppose we have data on a few employees at the Pawnee Parks and Recreation Department, shown in a data table below.

Name	Height (inches)	Weight (pounds)	Income (\$/month)
Leslie	62	115	4000
Ron	71	201	(Redacted)
April	66	119	2000

We can input the numeric data into a matrix in R.

```
parks.mat <- cbind(c(62,71,66),c(115,201,119),c(4000,NA,2000))
parks.mat # Make sure the data was entered correctly</pre>
```

```
## [,1] [,2] [,3]
## [1,] 62 115 4000
## [2,] 71 201 NA
## [3,] 66 119 2000
```

Question: Can you find at least two other ways to produce the same matrix of values in R?

While the numeric values from the dataset have been entered, the names of the employees and the names of the variables have been lost. To add names to the rows and columns of a two-dimensional object, we can use the rownames() and colnames() functions.

Applying the rownames() and colnames() functions to a two-dimensional object will retrieve (get), respectively, the current row and column names. If there are no names, the functions will output NULL.

```
rownames(parks.mat)
```

## NULL

```
colnames(parks.mat)
```

#### ## NULL

To set the names, we can create a character vector of names and use the assignment <- operator on the row and column names.

```
rownames(parks.mat) <- c("Leslie", "Ron", "April")
colnames(parks.mat) <- c("Height", "Weight", "Income")
parks.mat</pre>
```

```
## Height Weight Income
## Leslie 62 115 4000
## Ron 71 201 NA
## April 66 119 2000
```

Technically, these functions access the row and column name attributes of the input object. Setting names will add the dimnames attribute to the object.

```
attributes(parks.mat)
```

```
## $dim
## [1] 3 3
##
## $dimnames
## $dimnames[[1]]
## [1] "Leslie" "Ron" "April"
##
## $dimnames[[2]]
## [1] "Height" "Weight" "Income"
```

The attributes(parks.mat) object is an example of a list object in R. We will go over the syntax and functions for lists in a later chapter. For now, notice that the dimnames attribute has two components. The first component is the vector of row names, and the second component is the vector of column names.

Side Note: The dimnames() function can get and set both the row and column name attributes at once. The assignment input using dimnames() needs to be a list with two vector components.

```
dimnames(parks.mat) <- list(NULL, NULL) # Remove the row and column names
parks.mat
##
        [,1] [,2] [,3]
## [1,]
           62
               115 4000
## [2,]
          71
               201
                     NA
              119 2000
## [3,]
          66
# Add the same names as before
dimnames(parks.mat) <- list(c("Leslie", "Ron", "April"), c("Height", "Weight", "Income"))</pre>
dimnames(parks.mat)
```

```
## [[1]]
## [1] "Leslie" "Ron" "April"
##
## [[2]]
## [1] "Height" "Weight" "Income"
```

**Side Note**: There are a few other ways to add names to rows and columns. The matrix() function has an optional dimnames argument that allows us to add names directly when creating a matrix object. The syntax is the same as the dimnames() function.

```
matrix(1:9,nrow=3,ncol=3,dimnames=list(c("a","b","c"),c("A","B","C")))

## A B C
## a 1 4 7
## b 2 5 8
## c 3 6 9
```

The rbind() and cbind() functions allow us to name, respectively, each row or column by just typing the name of the row or column in quotation marks, as shown below. Setting names for the unnamed dimension would still need to be done separately.

```
rbind("a"=1:3,"b"=4:6,"c"=7:9)
     [,1] [,2] [,3]
## a
              2
                   3
              5
                   6
## b
## c
        7
              8
                   9
cbind("A"=1:3, "B"=4:6, "C"=7:9)
        A B C
## [1,] 1 4 7
## [2,] 2 5 8
## [3,] 3 6 9
```

# 3 Extracting Data From Two-Dimensional Objects

Recall that square brackets are used to extract specific parts of data from objects in R. Vectors are onedimensional objects, so a single number inside of square brackets extracts that element from a vector.

For two-dimensional objects, there are two index coordinates, corresponding to the row index and the column index. The index inside the square brackets needs to be an ordered pair, separated by a comma. For example, an index of [i,j] means to extract the entry in the ith row and jth column, also called the (i,j)th entry.

### 3.1 Numeric Indices

Leaving one value blank means to extract all the values in that dimension. Positive, negative, and fractional indices work the same way as they do for vectors. The row and column indices are independent of each other, so a positive index for one can be mixed with a negative index for the other.

```
B # Notice the row and column indices in the output
##
        [,1] [,2] [,3]
## [1,]
           1
                2
                      3
## [2,]
           4
                5
                      6
## [3,]
           7
B[2,1] # Extract the (2,1) element
## [1] 4
B[2,] # Extract the second row
## [1] 4 5 6
B[,3] # Extract the third column
```

```
## [1] 3 6 9
B[,-2] # Remove the second column
        [,1] [,2]
##
## [1,]
           1
                 3
## [2,]
           4
                 6
## [3,]
           7
                 9
B[-1,c(2,3)] # Remove the first row and extract the second and third columns
##
        [,1] [,2]
## [1,]
           5
                 6
## [2,]
           8
                 9
```

**Note**: Notice that when the resulting output is one-dimensional (i.e., a single row or a single column), the output object is a vector, not a one-dimensional matrix.

Caution: Using a single index [] instead of an index pair [,] will not throw a warning or an error for matrix objects. Since a matrix is stored as one long (column) vector, R will interpret the single index as a vector index and extract the values from the stored vector version of the matrix. The matrix values are stored in column-major storage order, meaning the values are stored in order from top to bottom down the first column, then top to bottom down the second column, etc. Using a single index for matrices is not recommended, since it can be hard to keep track of which entry in a two-dimensional array corresponds to a single index.

```
B[8] # Do not extract elements of matrices this way
## [1] 6
```

# 3.2 Logical Indices

Logical vectors can also be used for subsetting two-dimensional objects. The logical indices work the same way as they do for vectors. This can be a useful way to extract rows or columns that satisfy certain conditions.

```
# Which heights are above 65 inches?
tall.index <- parks.mat[,1] > 65
# Extract only the rows/observations for people who are taller than 65 inches
parks.mat[tall.index,]

## Height Weight Income
## Ron 71 201 NA
## April 66 119 2000
```

Question: How can we use the tall.index vector to extract only the rows/observations for people in the data who are at most 65 inches tall?

## 3.3 Named (Character) Indices

If the rows or columns of a two-dimensional object are named, we can use the name as an index. Row names can only be used as a row index, and column names can only be used as a column index.

```
# Extract the row of data for Leslie
parks.mat["Leslie",]

## Height Weight Income
## 62 115 4000
```

```
# Extract the column of data for Income
parks.mat[,"Income"]
## Leslie
                  April
             Ron
##
     4000
                    2000
# Extract the height of Ron
parks.mat["Ron","Height"]
## [1] 71
# Extract the weight of Leslie and April
parks.mat[c("Leslie", "April"), "Weight"]
## Leslie
           April
##
      115
             119
```

**Note**: Notice that we do not need to know the numeric index for the observations or variables. Using names as indices can be useful if the object has many rows or columns and the numeric index is not as easy to use.

# 4 Matrix Operations

### 4.1 Entrywise Arithmetic Operations

Because matrices are stored as vectors, the usual arithmetic operations (+, -, \*, /, ^, %%, %/%) that apply to numeric vectors will also work for numeric matrices. Just like for vectors, arithmetic operations will be applied entrywise. However, for arithmetic operations between matrices to make sense, the matrices need to be conformable. We can only apply the usual arithmetic operators to matrices with the same dimensions (same number of rows and columns). The corresponding entries of the two matrices will be operated together.

```
A + 10 # Add 10 to every entry in A
##
        [,1] [,2] [,3]
## [1,]
          11
                13
                     15
## [2,]
          12
                14
                     16
A^2 # Square each entry in A
        [,1] [,2] [,3]
##
## [1,]
           1
                 9
                     25
## [2,]
                16
                     36
C <- matrix(1:3,nrow=2,ncol=3) # Construct a matrix C
C
##
        [,1] [,2] [,3]
## [1,]
                 3
           1
## [2,]
           2
A + C # Add A and C
##
        [,1] [,2] [,3]
           2
                 6
                      7
## [1,]
## [2,]
           4
                 5
A ^ C # Exponentiate A by C
        [,1] [,2] [,3]
```

```
## [1,] 1 27 25
## [2,] 4 4 216
```

### A \* C # Multiply A and C (entrywise)

```
## [,1] [,2] [,3]
## [1,] 1 9 10
## [2,] 4 4 18
```

If we try to apply the operators to matrices that are not conformable, R will throw an error.

```
A * B # A and B are not conformable
```

## Error in A \* B: non-conformable arrays

### 4.2 Matrix Multiplication

Caution: The entrywise multiplication performed using the regular multiplication \* operator is not the way matrices are multiplied together in linear algebra.

For **matrix multiplication**, two matrices are conformable if the columns of rows in the left matrix has to be the same as the number of rows in the right matrix. If A is an  $m \times n$  matrix and B is an  $n \times p$  matrix, then A and B are conformable, and AB will be an  $m \times p$  matrix.

For example, let A be a  $2 \times 3$  matrix and B be a  $3 \times 2$  matrix, denoted by

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \\ b_{31} & b_{32} \end{bmatrix}.$$

The matrix product AB is the  $2 \times 2$  matrix given by

$$AB = \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} + a_{13}b_{31} & a_{11}b_{12} + a_{12}b_{22} + a_{13}b_{32} \\ a_{21}b_{11} + a_{22}b_{21} + a_{23}b_{31} & a_{21}b_{12} + a_{22}b_{22} + a_{23}b_{32} \end{bmatrix}.$$

Written generally, if  $A = [a_{ik}]_{m \times n}$  and  $B = [b_{kj}]_{n \times p}$ , then  $AB = \left[\sum_{k=1}^{n} a_{ik} b_{kj}\right]_{m \times p}$ . The (i, j)th entry of the product AB is the **dot product** of the ith row of A with the jth column of B.

Matrix multiplication is performed using the **%\*%** operator.

```
A %*% C # Non-conformable error
```

## Error in A %\*% C: non-conformable arguments

A %\*% B # Matrix multiplication of A and B

Notice that matrix  $\mathtt{A}$  and  $\mathtt{C}$  are conformable for entrywise multiplication but are not conformable for matrix multiplication, while  $\mathtt{A}$  and  $\mathtt{B}$  are conformable for matrix multiplication but not for entrywise.

### 4.3 The Identity Matrix

The **identity matrix** of size n, denoted by  $I_n$  (or simply I if the dimension is implicit) is an  $n \times n$  square matrix with 1's in the main diagonal entries and 0 everywhere else. Formally, the (i, j)th entry of  $I_n$  is 1 if i = j and 0 if  $i \neq j$ , written mathematically as

$$[I_n]_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j. \end{cases}$$

For example:

$$I_1 = \begin{bmatrix} 1 \end{bmatrix}, \quad I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \dots$$

The identity matrix essentially plays the role of the number 1 for matrix multiplication. That is, for any  $n \times n$  matrix A, a property of matrix multiplication is that  $AI_n = I_n A = A$ .

The diag() function has two main functionalities. By inputting a number, the diag() function will generate an identity matrix of that size.

```
diag(4) # Create a 4x4 identity matrix
```

```
## [,1] [,2] [,3] [,4]
## [1,] 1 0 0 0
## [2,] 0 1 0 0
## [3,] 0 0 1 0
## [4,] 0 0 0 1
```

By inputting a vector, the diag() function will generate a diagonal matrix (the only nonzero entries are along the diagonal) with the vector values along the diagonal.

```
diag(c(1,2,3)) # Create a diagonal matrix with 1, 2, 3 along the diagonal
```

```
## [,1] [,2] [,3]
## [1,] 1 0 0
## [2,] 0 2 0
## [3,] 0 0 3
```

The nrow and ncol arguments can also be used to specify the dimensions for a rectangular (non-square) matrix.

```
diag(c(1,2,3),nrow=3,ncol=4)
```

```
## [,1] [,2] [,3] [,4]
## [1,] 1 0 0 0
## [2,] 0 2 0 0
## [3,] 0 0 3 0
```

### 4.4 The Inverse of a Matrix

The **inverse** of an  $n \times n$  square matrix A is the unique  $n \times n$  matrix, denoted by  $A^{-1}$ , such that

$$AA^{-1} = A^{-1}A = I_n.$$

The inverse of a matrix is analogous to the reciprocal (also called the *multiplicative inverse*) for numbers. For example, the reciprocal of 2 is  $2^{-1} = \frac{1}{2}$ , since  $2 \times \frac{1}{2} = \frac{1}{2} \times 2 = 1$ .

The function solve() computes the inverse of the inputted matrix. For example:

```
M <- matrix(c(1,4,2,1),nrow=2,ncol=2) # Create a matrix M
##
        [,1] [,2]
## [1,]
           1
                2
## [2,]
           4
                1
M.inv <- solve(M) # Compute the inverse of M
M.inv
##
              [,1]
                          [,2]
## [1,] -0.1428571 0.2857143
## [2,] 0.5714286 -0.1428571
# Verify that M.inv is the inverse of M
M %*% M.inv
        [,1] [,2]
##
## [1,]
           1
                0
## [2,]
M.inv %*% M
##
        [,1] [,2]
## [1,]
           1
## [2,]
           0
                1
```

Caution: The inverse of a (square) matrix does not always exist. A matrix that has an inverse is called invertible or nonsingular. There are conditions in linear algebra for when an inverse exists. A matrix that does not have an inverse is called **singular**. If we try to invert a singular matrix, R will throw an error.

```
solve(B)
```

```
## Error in solve.default(B): system is computationally singular: reciprocal condition
## number = 2.59052e-18
```