# Chapter 4 Matrices, Data Frames and Lists 

Math 3210

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## Outlines

- Matrices
- Arrays
- Data Frames
- Lists


## Data Structure in R

- Vectors
- Factors
- Matrices
- Arrays
- Data Frames
- Lists


## Constructing Matrix objects

A matrix is a rectangular way of storing data. You can simply view it as a way to store data. Matrices can be constructed using the functions matrix().
m<-matrix(1:6,nrow=2,ncol=3)
m

|  | $[, 1]$ | $[, 2]$ | $[, 3]$ |
| :--- | ---: | ---: | ---: |
| $[1]$, | 1 | 3 | 5 |
| $[2]$, | 2 | 4 | 6 |

Matrices are constructed column-wise, so entries can be thought of starting in the "upper left" cornere and running down the columns. If you need the matrix to be filled by rows, then add the byrow=TRUE argument.

We can then access elements using two indice. For example, $m[1,2]$
[1] 3
m[1,2:3]
[1] 35

Whole rows or columns of matrices may be selected by leaving one index blank:
$\mathrm{m}[1$,$] \#display the first row of m$
[1] 135
$m[, 1]$ \#display the first column of $m$
[1] 12

To show the dimension of the matrix, we can use $\operatorname{dim}()$.
$\operatorname{dim}(m)$
[1] 23

To assign column and row names, we can use

```
rownames(m)<-c("obs1", "obs2")
colnames (m)<-LETTERS [1:3]
```

m

```
    A B C
obs1 1 3 5
obs2 2 4 6
```

Matrices can also be created by cbind(), or rbind().

```
x<-1:3
y<-10:12
z<--5:-3
cbind(x,y,z)
```

[1, $\begin{array}{lll}x & y & z\end{array}$
$[1] \quad 110-$,
$[2] \quad 2111-$,
$[3] \quad 312-$,
rbind ( $\mathrm{x}, \mathrm{y}, \mathrm{z}$ )

|  | $[, 1]$ | $[, 2]$ | $[, 3]$ |
| :--- | ---: | ---: | ---: |
| x | 1 | 2 | 3 |
| y | 10 | 11 | 12 |
| z | -5 | -4 | -3 |

A more general way to store data is in an array. Arrays have multiple indices, and created using the array() function:

```
a<-array(1:24,c(3,4,2))
a
```

, , 1

|  | $[, 1]$ | $[, 2]$ | $[, 3]$ | $[, 4]$ |
| :--- | ---: | ---: | ---: | ---: |
| $[1]$, | 1 | 4 | 7 | 10 |
| $[2]$, | 2 | 5 | 8 | 11 |
| $[3]$, | 3 | 6 | 9 | 12 |

, , 2

|  | $[, 1]$ | $[, 2]$ | $[, 3]$ | $[, 4]$ |
| :--- | ---: | ---: | ---: | ---: |
| $[1]$, | 13 | 16 | 19 | 22 |
| $[2]$, | 14 | 17 | 20 | 23 |
| $[3]$, | 15 | 18 | 21 | 24 |

## Data Frames

- Data frames are used to store tabular data in R. They are an important type of object in R and are used in a variety of statistical modeling applications.
- Most data sets are stored in R as data frames. They are like matrices, but with the columns having their own names.
- Unlike matrices, data frames can store different classes of objects in each column. Matrices must have every element be the same class.


## Example:

The data frame women contains the average weights (in pounds) of American women aged 30 to 39 of particular heights (in inches):
women

|  | height | weight |
| :--- | ---: | ---: |
| 1 | 58 | 115 |
| 2 | 59 | 117 |
| 3 | 60 | 120 |
| 4 | 61 | 123 |
| 5 | 62 | 126 |
| 6 | 63 | 129 |
| 7 | 64 | 132 |
| 8 | 65 | 135 |
| 9 | 66 | 139 |
| 10 | 67 | 142 |
| 11 | 68 | 146 |
| 12 | 69 | 150 |
| 13 | 70 | 154 |
| 14 | 71 | 159 |
| 15 | 72 | 164 |

Sometimes, the data frames can be very large, it is not a good idea to display the entire data set. To obtain the basic information of the data set, we can use

```
nrow(women) #show the number of rows
[1] 15
ncol(women) #show the number of columns
[1] 2
dim(women) #show both dimensions
[1] 15 2
rownames(women) #show the row names
    [1] "1" "2" "3" "4" "5" "6" "7" "8" "9" "10" "11" "12" "13" "
[15] "15"
colnames(women) #show the column names
```

We can use the $\boldsymbol{s t r}()$ function to get summary information for structure. It works with almost any R object and it is often a quick way to find out what you are working with.

```
str(women)
'data.frame': }15\mathrm{ obs. of 2 variables:
    $ height: num 58 59 60 61 62 63 64 65 66 67 ...
    $ weight: num 115 117 120 123 126 129 132 135 139 142 ...
```

To summarize the data numerically,

```
summary(women)
```

```
        height weight
    Min. :58.0 Min. :115.0
    1st Qu.:61.5 1st Qu.:124.5
    Median :65.0 Median :135.0
    Mean :65.0 Mean :136.7
    3rd Qu.:68.5 3rd Qu.:148.0
    Max. :72.0 Max. :164.0
colMeans(women) # compute the mean for each column
    height weight
    65.0000 136.7333
rowMeans(women) # compute the mean for each row
    [1] 86.5 88.0 90.0 92.0 94.0
[12] 109.5 112.0 115.0 118.0
```


## The apply() function

The apply( ) function allows you to apply a function to each row or to each column of data.

```
apply(X, MARGIN, FUN, ...) # applies a function to each row or column.
```

\#\# X: a data frame or matrix
\#\# MARGIN: 1 indicates rows, 2 indicates columns
\#\# FUN: the function to be applied to each row or column
apply( X= trees, MARGIN= 2, FUN= mean) \# same output as colMeans
Girth Height Volume
13.2483976 .0000030 .17097
apply ( X= trees, MARGIN= 2, FUN= median)
Girth Height Volume
$12.9 \quad 76.0 \quad 24.2$

To summarize the data graphically,
plot(weight~height, data=women)


## Indexing of data frames

We can extract elements from data frames using similar syntax to what was used with matrices. For example, women [7,2]
[1] 132
women [3,]

```
    height weight
3 60 120
```

women [4:7,1]
[1] 61626364

## The \$ operator

```
Data frame columns can also be addressed using their namems using the \$ operator. For example, the weight column can be extracted as follows:
women\$weight
[1] \(1 \begin{array}{llllllllllllll}115 & 117 & 120 & 123 & 126 & 129 & 132 & 135 & 139 & 142 & 146 & 150 & 154 & 159\end{array} 164\)
women\$height
[1] \(\begin{array}{llllllllllllllll}58 & 59 & 60 & 61 & 62 & 63 & 64 & 65 & 66 & 67 & 68 & 69 & 70 & 71 & 72\end{array}\)
```


## Conditional Selection

```
Sometimes, it can be useful to extract all data for cases that satisfy some criterion. For example, we can extract all heights for which the weights exceed 140 using
women$height [women$weight>140]
[1] 67 68 69 70 71 72
women$weight>140
    [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE
[12] TRUE TRUE TRUE TRUE
which(women$weight>140) #show the women whose weights exceed 140
[1] 10}1011 12 13 14 15,
```


## The with() function

The with() function allows us to access columns of a data frame directly without using the $\$$. For example, we can divide the weights by the heights in the women data frame using

```
with(women,weight/height) #use the with() function
```

    [1] 1.9827591 .9830512 .0000002 .0163932 .0322582 .0476192 .062500
    [8] 2.0769232 .1060612 .1194032 .1470592 .1739132 .2000002 .239437
    [15] 2.277778
women\$weight/women\$height \#without the with() function
[1] 1.9827591 .9830512 .0000002 .0163932 .0322582 .0476192 .062500
[8] 2.0769232 .1060612 .1194032 .1470592 .1739132 .2000002 .239437
[15] 2.277778

## Constructing data frames

Use the data.frame() function to construct data frames from vectors that already exist in your workspace:
$x<-1: 5$
$y<-7: 3$
z<-11:15
xyzdata<-data.frame( $x, y, z$ )
xyzdata

|  | $x$ | $y$ | $z$ |
| ---: | ---: | ---: | ---: |
| 1 | 1 | 7 | 11 |
| 2 | 2 | 6 | 12 |
| 3 | 3 | 5 | 13 |
| 4 | 4 | 4 | 14 |
| 5 | 5 | 3 | 15 |

Alternatively, we can use
xyzdata<-data.frame ( $x=1: 5, y=7: 3, z=11: 15$ )

## Non-numeric columns in data frames

Columns of data frames can be of different types. For example, the built in data frame 'chickwts' has a numeric column and a factor. chickwts

To view the basic information about the data frame 'chickwts' str (chickwts)

```
'data.frame': 71 obs. of 2 variables:
    $ weight: num 179 160 136 227 217 168 108 124 143 140 ...
    $ feed : Factor w/ 6 levels "casein","horsebean",..: 2 2 2 2 2 2 2 2
summary(chickwts)
```

| weight |  | feed |  |
| :--- | :--- | :--- | :---: |
| Min. | $: 108.0$ |  |  |
| casein | $: 12$ |  |  |
| 1st Qu.:204.5 | horsebean:10 |  |  |
| Median $: 258.0$ | linseed $: 12$ |  |  |
| Mean | $: 261.3$ |  |  |
| meatmeal $: 11$ |  |  |  |
| 3rd Qu. $: 323.5$ |  | soybean $: 14$ |  |
| Max. | $: 423.0$ |  |  |
| sunflower $: 12$ |  |  |  |

## Example:

Consider the following data that might be used as a baseline in an obesity study:

```
gender<-c("M","M","F","F","F")
weight<-c(73,68,52,69,64)
obesitystudy<-data.frame(gender,weight)
obesitystudy
```

    gender weight
    1 M 73
2 M 68
3 F 52
$4 \quad$ F 69
5 F 64

## Practice: Changing variable names

obesitystudy\$gender
[1] M M F F F
Levels: F M

Obviously, the character vector 'gender' is convereted to a factor in the data frame. Suppose we wish to globally change F to Female in the data frame.

```
as.integer(obesitystudy$gender)
```

[1] 222111
levels(obesitystudy\$gender) [1]<-"Female" \#F is the 1st level
obesitystudy\$gender \#Check the new data frame
[1] M M Female Female Female

Levels: Female M

## Grouped data and data frames

- The natural way of storing grouped data in a data frame is to have the data themselves in one vector and parallel to that have a factor telling which data are from which group.
- However, sometimes it is desirable to have data in a separate vector for each group.
- For example, suppose we want to extract the weights of chickens in the soybean group.
soybeangrp<-chickwts\$weight [chickwts\$feed=="soybean"]
soybeangrp
[1] 243230248327329250193271316267199171158248


## The split() function

Alternatively, you can use the split() function, which generates a list of vectors according to a grouping.

```
split(chickwts$weight, chickwts$feed)
```

\$casein
[1] $\begin{array}{lllllllllllllllllllll}368 & 390 & 379 & 260 & 404 & 318 & 352 & 359 & 216 & 222 & 283 & 332\end{array}$
\$horsebean
[1] $17 \begin{array}{lllllllll}179 & 160 & 136 & 227 & 217 & 168 & 108 & 124 & 143 \\ 140\end{array}$
\$linseed
[1] $\begin{array}{llllllllllll}309 & 229 & 181 & 141 & 260 & 203 & 148 & 169 & 213 & 257 & 244 & 271\end{array}$
\$meatmeal
[1] $\begin{array}{lllllllllll}325 & 257 & 303 & 315 & 380 & 153 & 263 & 242 & 206 & 344 & 258\end{array}$
\$soybean
[1] $243230248327329250193 \quad 271 \quad 316$
\$sunflower
[1] $423 \quad 340 \quad 392339 \quad 341 \quad 226$

## Lists

Data frames are actually a special kind of list. List in R can contain any other objects. The list() function is one way of organizing multiple pieces of output from functions.

## Example 1:

```
a <- matrix( c( 1, 2, 3, 4 ), nrow= 2, ncol= 2 )
b <- c(1, 3, 7)
c <- "hello!"
d <- c(TRUE, FALSE, FALSE, TRUE)
z <- list( "FirstItem"=a, "SecItem"=b, "ThirdItem"=c, "FourthItem"=d)
z
\$FirstItem
\begin{tabular}{lrr} 
& {\([, 1]\)} & {\([, 2]\)} \\
{\([1]\),} & 1 & 3 \\
{\([2]\),} & 2 & 4
\end{tabular}
\$SecItem
[1] 137
\$ThirdItem
[1] "hello!"
\$FourthItem
[1] TRUE FALSE FALSE TRUE
```


## Working with Lists

You can see the names of the objects in a list using the names() function, and extract parts of it:

```
names(z) #Print the names of the objects in z
[1] "FirstItem" "SecItem" "ThirdItem" "FourthItem"
z$SecItem # Print the FirstItem of z
[1] 1 3 7
z[[2]] #You can use this when the elements in the list don't have names
[1] 1 3 7
```


## Example 2:

```
\(\mathrm{x}<-\mathrm{c}(2,7,8,6)\)
\(\mathrm{y}<-\mathrm{c}(1,3,8,9)\)
z <- c(4, 2, 7, 3)
G <- list( x, y, z )
G \# elements don't have names! Try names (G)
```

[[1]]
[1] 2786
[[2]]
[1] 1389
[[3]]
[1] 4273

G[[3]]
[1] 4273

## The lapply() function

There are several functions which make working with lists easy. Two of them are lapply() and vapply(). The lapply() function "applies" another functin to every element of a list and returns the results in a new list. The general form is:
lapply ( X, FUN )
\#\# X: a list
\#\# FUN: the function to be applied to each element of X
lapply (G,mean)
[[1]]
[1] 5.75
[[2]]
[1] 5.25
[[3]]
[1] 4
lapply(G,fivenum) \#returns the five numbers for each list element

## The vapply() function

In the previous example, it might be more convenient to have the results in a vector. The vapply() does the same thing as lapply( ) but return the result of the function in vector form. The general form is:
vapply( X, FUN, FUN.VALUE ) \#applies a function to each list element. \#\# X: a list
\#\# FUN: the function to be applied to each element of $X$
\#\# FUN.VALUE: An example of the output
vapply(X= G, FUN= mean, FUN.VALUE= 1) \# the '1' serves as an example
[1] 5.755 .254 .00

If we expect more than a single value, the results will be organized into a matrix, e.g.

```
vapply(X= G, FUN= summary, numeric(6) )
```

|  | $[, 1]$ | $[, 2]$ | $[, 3]$ |
| :--- | :--- | :--- | :--- |
|  |  |  |  |
| Min. | 2.00 | 1.00 | 2.00 |
| 1st Qu. | 5.00 | 2.50 | 2.75 |
| Median | 6.50 | 5.50 | 3.50 |
| Mean | 5.75 | 5.25 | 4.00 |
| 3rd Qu. | 7.25 | 8.25 | 4.75 |
| Max. | 8.00 | 9.00 | 7.00 |

vapply(X= G, FUN= summary, FUN.VALUE= $c(1,1,1,1,1,1)$ )

|  | $\left[\begin{array}{lll} & {[1]} & {[, 2]}\end{array}[, 3]\right.$ |  |  |
| :--- | :--- | :--- | :--- |
| Min. | 2.00 | 1.00 | 2.00 |
| 1st Qu. | 5.00 | 2.50 | 2.75 |
| Median | 6.50 | 5.50 | 3.50 |
| Mean | 5.75 | 5.25 | 4.00 |
| 3rd Qu. | 7.25 | 8.25 | 4.75 |
| Max. | 8.00 | 9.00 | 7.00 |

## More examples:

```
xvec <- c( 2, 4, 8)
yvec <- c(5, 4, 10, 44)
range(xvec) # no element names
```

[1] 28
vecsList <- list( xvec, yvec)
vapply(X= vecsList, FUN= range, FUN.VALUE= c( 1, 1 ) )

|  | $[, 1]$ | $[, 2]$ |
| :--- | ---: | ---: |
| $[1]$, | 2 | 4 |
| $[2]$, | 8 | 44 |

vapply(X= vecsList, FUN= range, numeric(2) )

|  | $[, 1]$ | $[, 2]$ |
| :--- | ---: | ---: |
| $[1]$, | 2 | 4 |
| $[2]$, | 8 | 44 |

Sometimes, we need to add names to the outputs.
vapply ( vecsList , range, c( Min. = 1, Max. = 0) )

|  | $[, 1]$ | $[, 2]$ |
| :--- | ---: | ---: |
| Min. | 2 | 4 |
| Max. | 8 | 44 |

